

Wavelets at the Galactic Center

Sam McDermott

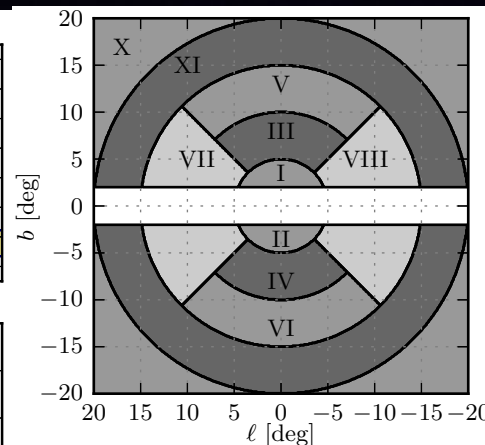
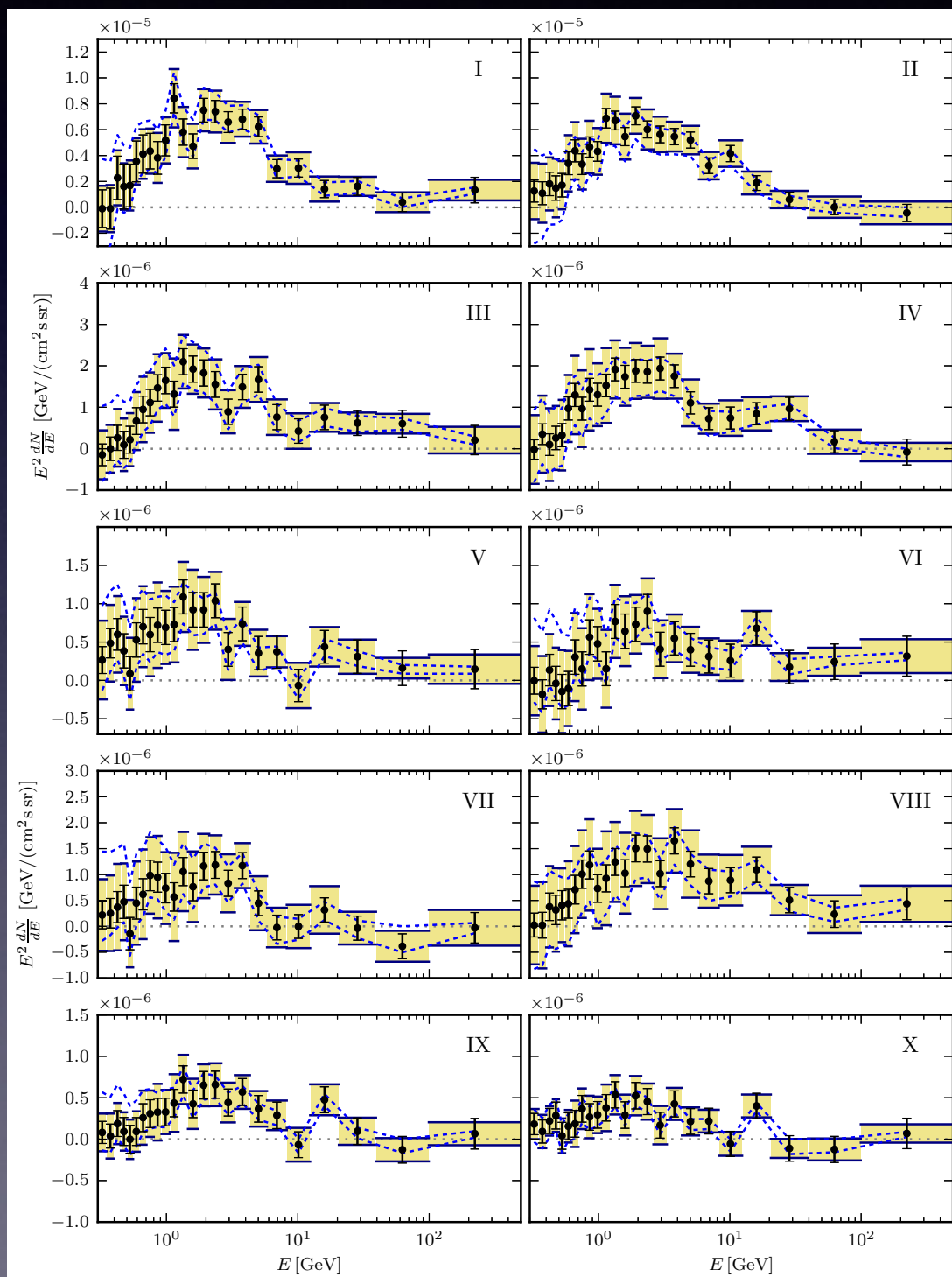
Based on:

SDM, I. Cholis, P. Fox, S. K. Lee
(preliminary / in progress)

Brookhaven Forum, 10/08/15



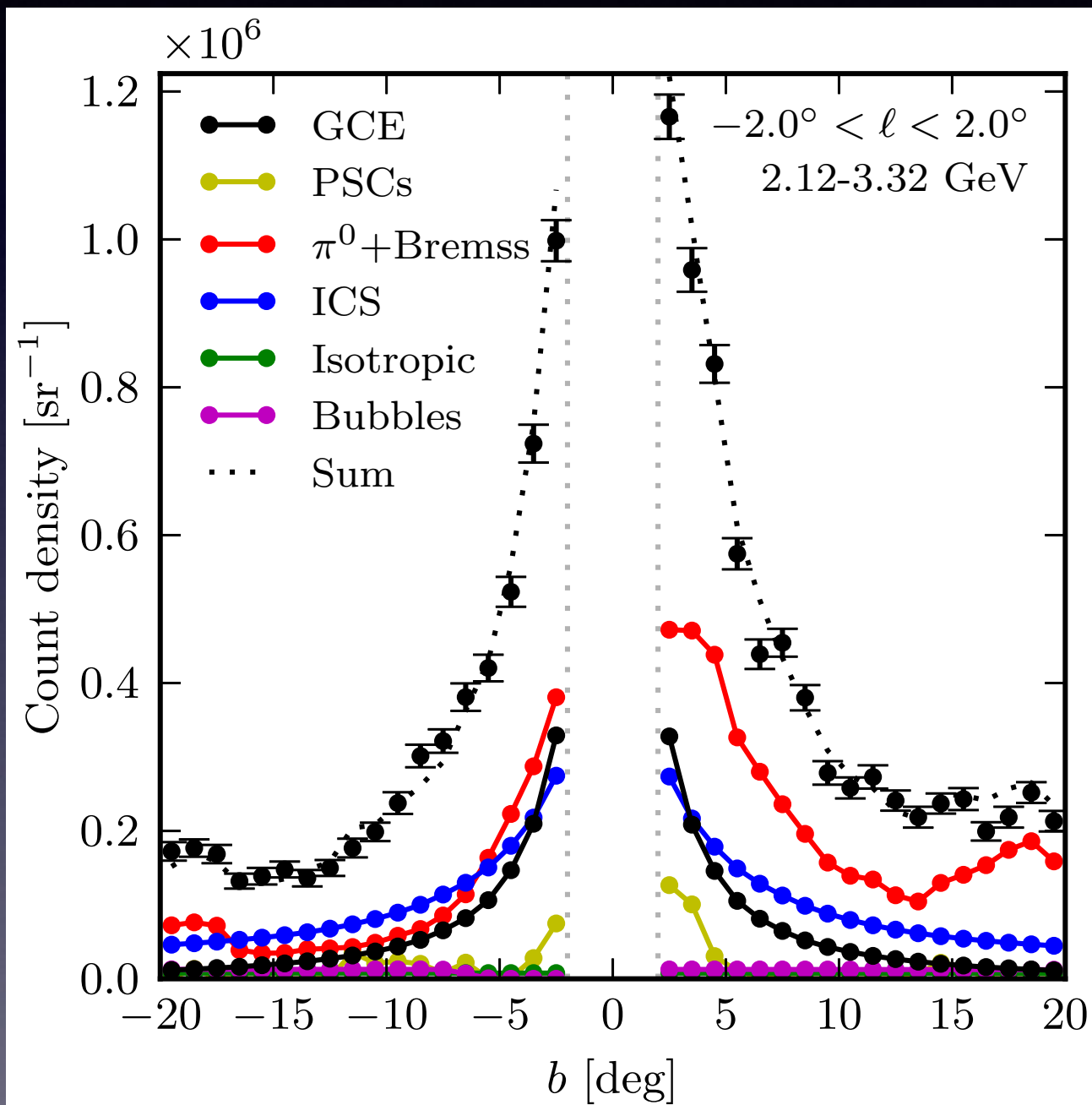
Excess photons



Calore, Cholis, Weniger
1409.0042

$E_{\gamma, \text{max}} \sim 2 \text{ GeV}$,
robust to changes
in diffuse template

Total Normalization



at energies of
interest, $\sim \mathcal{O}(30\%)$
of total flux

1409.0042

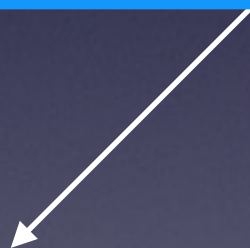
“Introduction”

existence of a significant new gamma-ray emission from Galactic Center is pretty robust, but...

- ... diffuse templates house large, energy-dependent uncertainties
- ... serious caution and healthy skepticism are required when interpreting as BSM physics
- ... a few opportunities so far that “could have been convincing” (either way) have not panned out

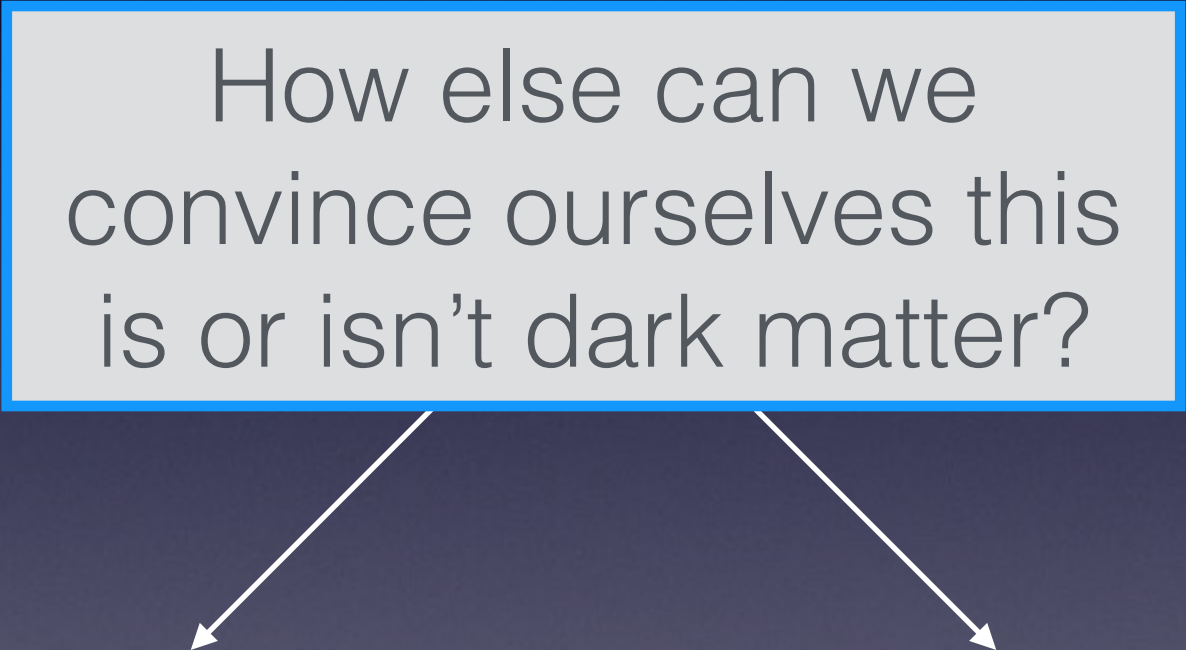
How else can we
convince ourselves this
is or isn't dark matter?

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Particle physics ideas

How else can we
convince ourselves this
is or isn't dark matter?



```
graph TD; A[How else can we convince ourselves this is or isn't dark matter?] --> B[Particle physics ideas]; A --> C[New observational ideas];
```

Particle physics ideas

New observational ideas

Current Technique

Test assumption of dark matter annihilation:

- statistical discrimination (χ^2 test) between fits with and without signal template
- fits with template do better

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...but what if there is a totally different shape on the sky that was not adequately tested?

Current Technique

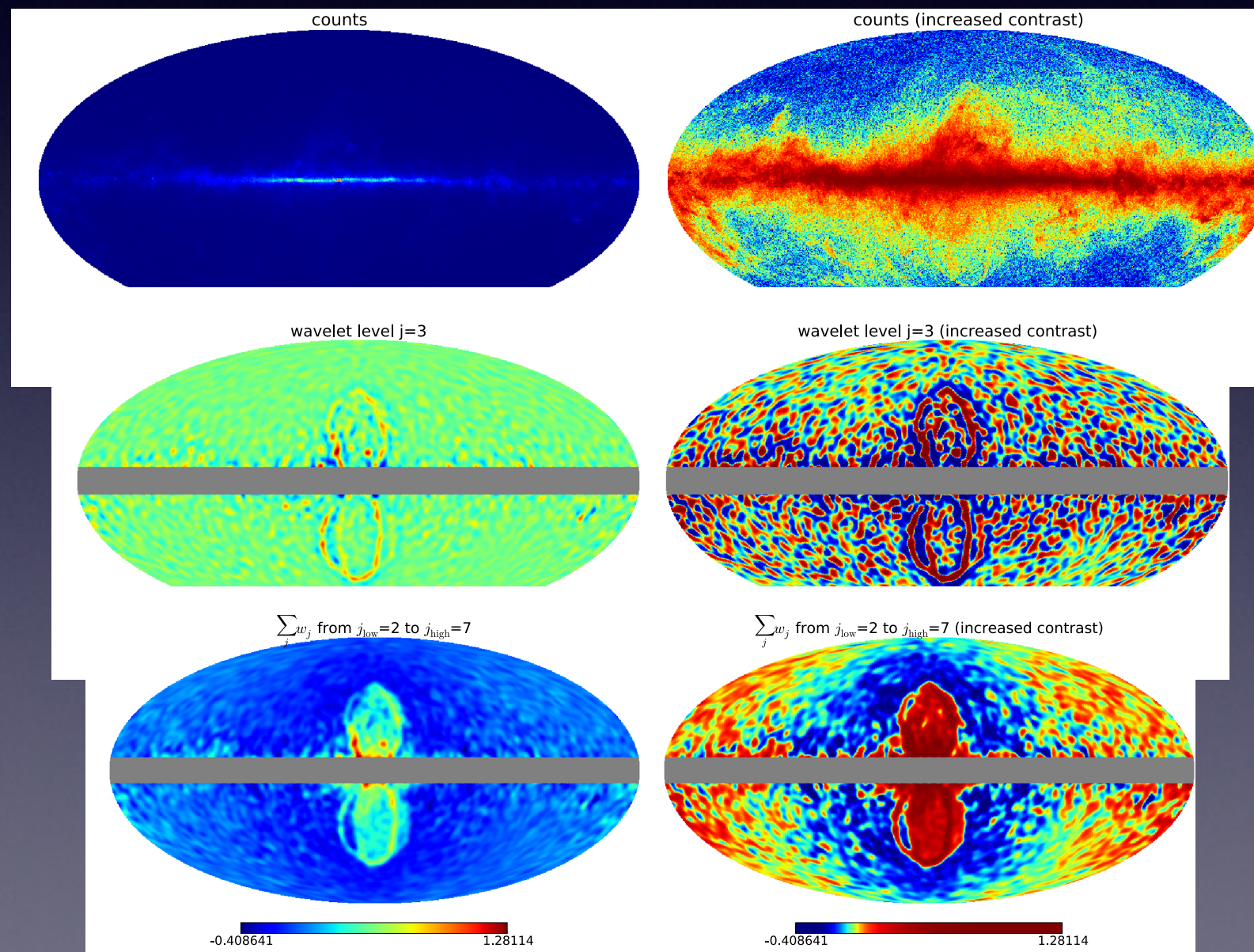
Test assumption of dark matter annihilation:

- It would be nice to find evidence without making this assumption!
-

...but what if there is a totally different shape on the sky that was not adequately tested?

Rudimentary Image Processing, with Wavelets

in development with Paddy Fox,
Ilias Cholis, and Samuel K Lee



Wavelets

Allow analysis sensitive to both location and scale

Used for a wide variety of industrial and academic applications:

- image compression (JPEG-2000)
- fast astrophysical signal identification
- cochlear transforms (mimic hearing)
- image denoising
- jets (this is still in its infancy...)
- etc.**

What are wavelets?

wavelet coefficients

original signal

$$W(a, b) = \frac{1}{\sqrt{a}} \int f(x) \psi^* \left(\frac{x - b}{a} \right) dx$$

scale position mother wavelet
(different choices)

$$\int \psi(x) dx = 0$$
$$\int |\psi(x)|^2 dx = 1$$

$$\psi(x) \in \mathbb{L}^2(\mathbb{R}) \text{ and}$$

$$\frac{1}{\sqrt{a}} \psi \left(\frac{x - b}{a} \right) \in \mathbb{L}^2(\mathbb{R})$$

$$\text{for } a, b \in \mathbb{Z}$$

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$$W(a, b) = \frac{1}{\sqrt{a}} \int f(x) \psi^* \left(\frac{x - b}{a} \right) dx$$

scale

How (and why)
do they work?

other wavelet
choices)

$$\int \psi(x) dx = 0$$

$$\int |\psi(x)|^2 dx = 1$$

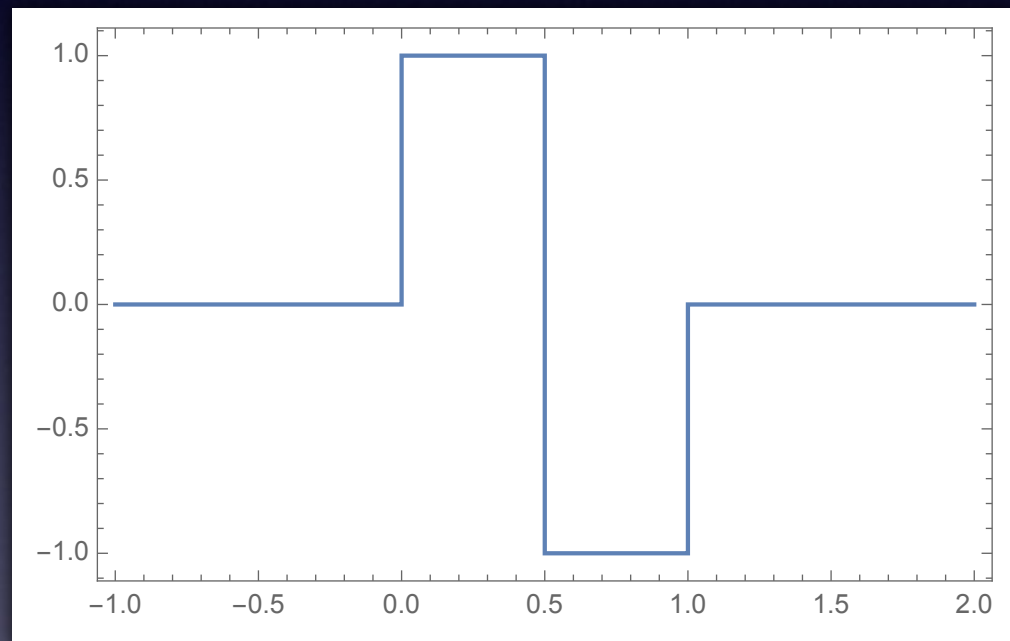
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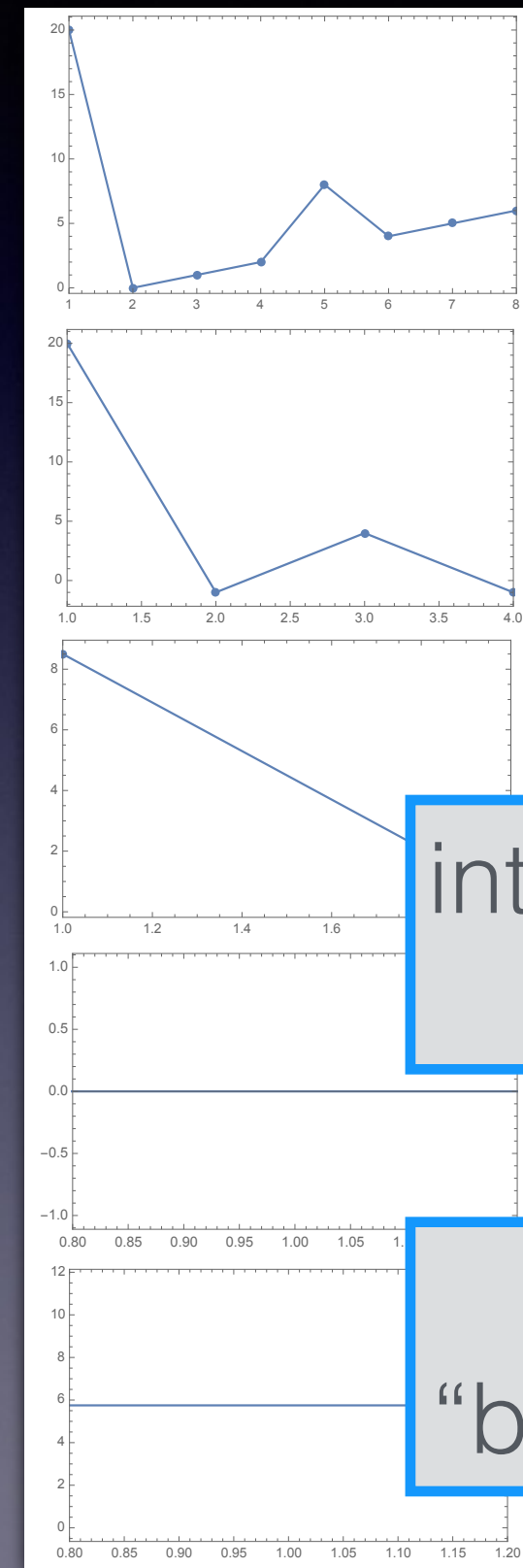
$$\text{for } a, b \in \mathbb{Z}$$

Haar

The Haar wavelet:



differences adjacent
cells (or pixels, etc.)



signal

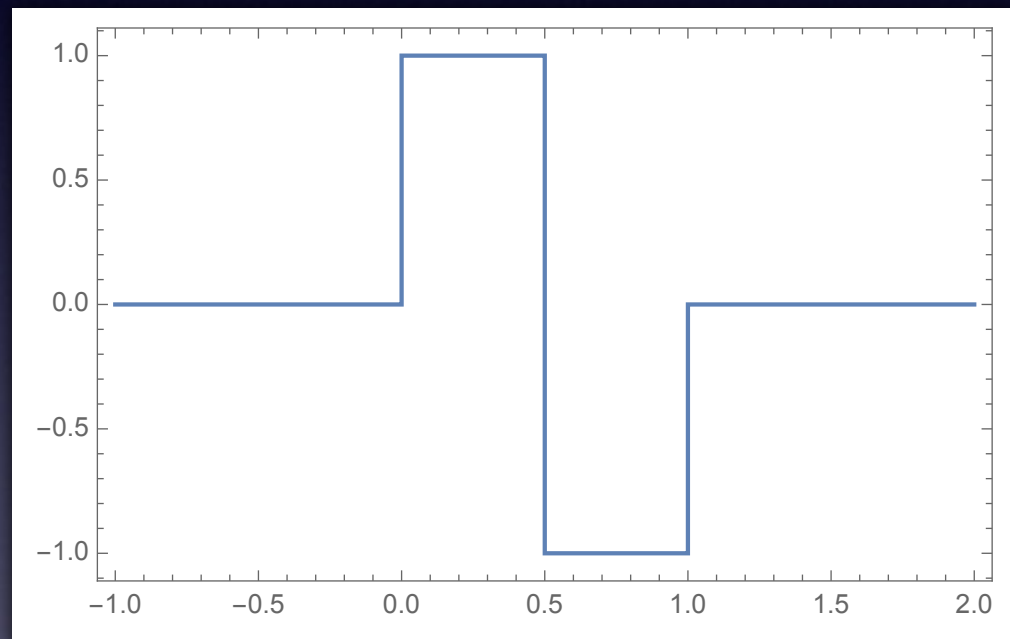
small
scale

intermediate
scales

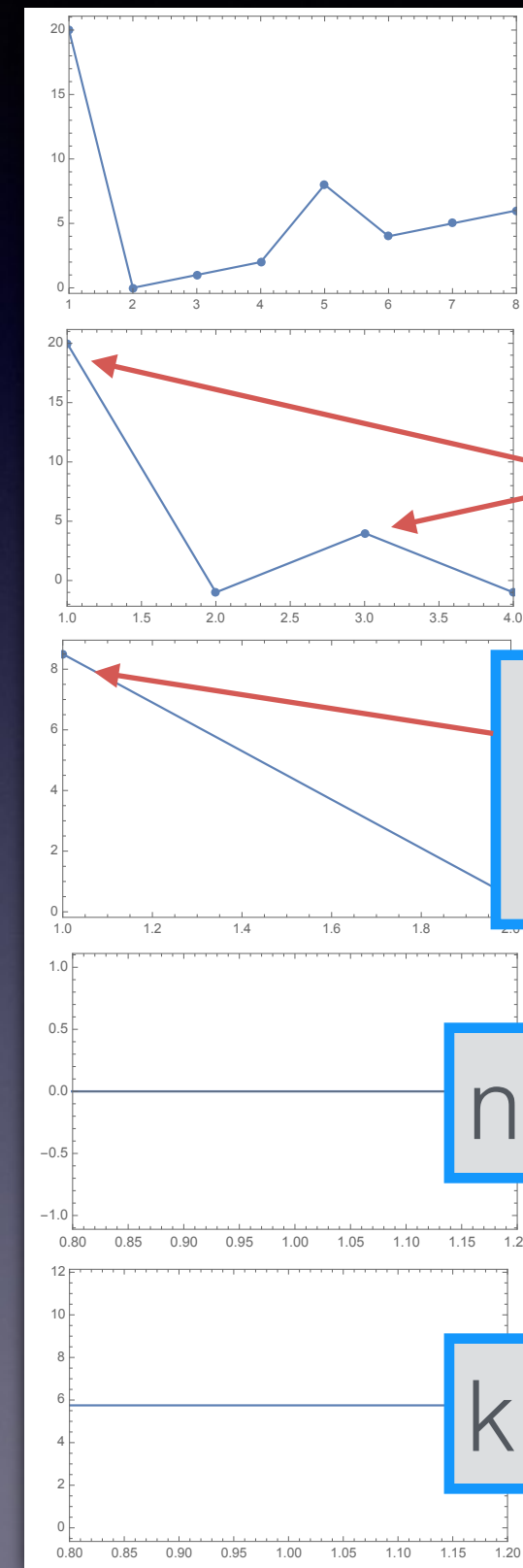
overall
"brightness"

Haar

The Haar wavelet:



differences adjacent
cells (or pixels, etc.)



signal

edges!

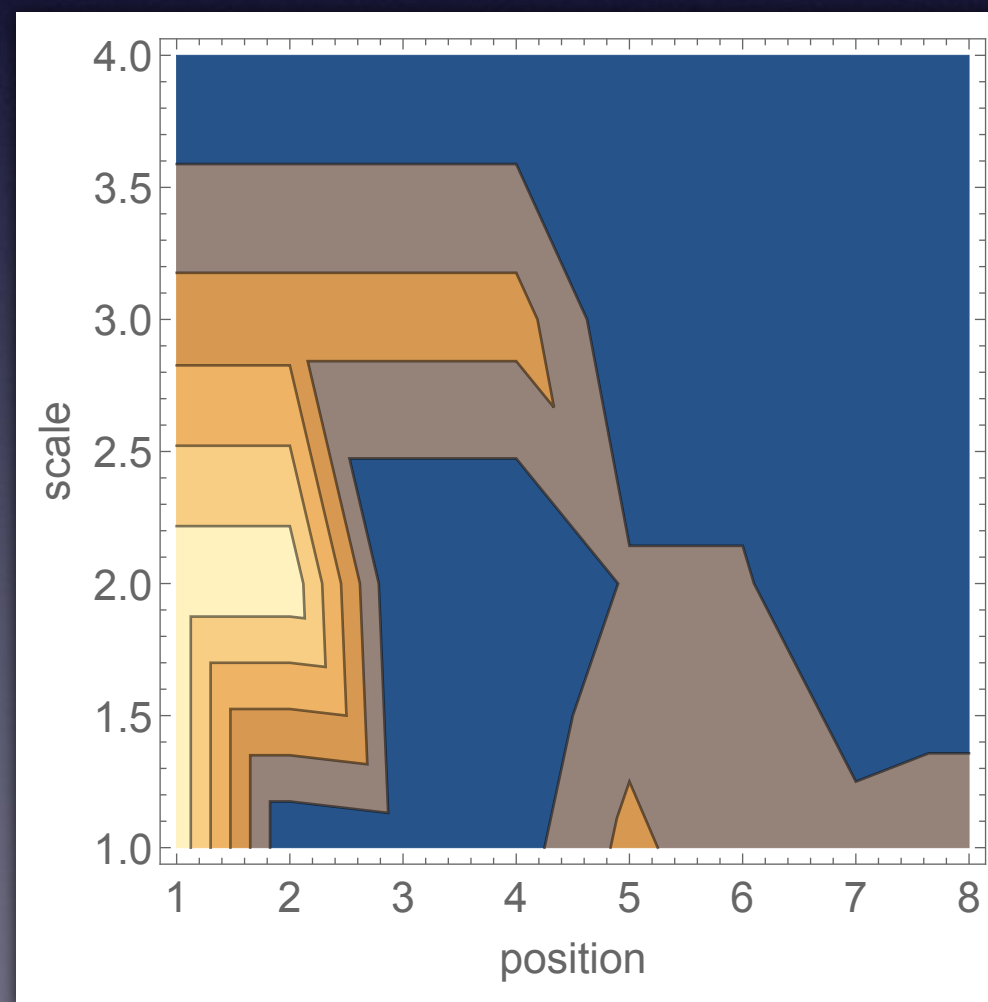
this edge
is bigger!

no structure

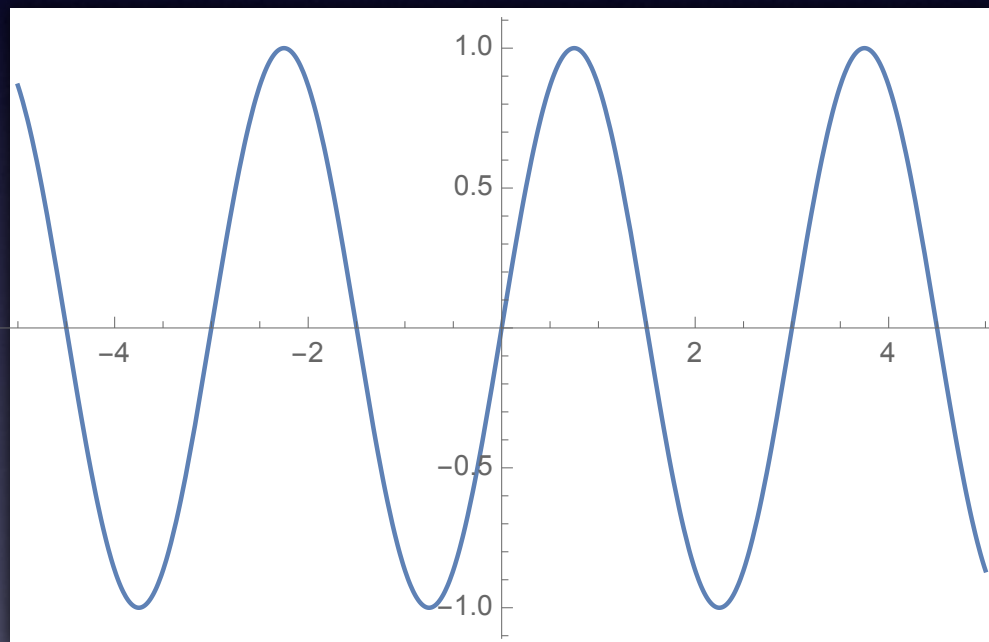
kinda bright

Alternate Form

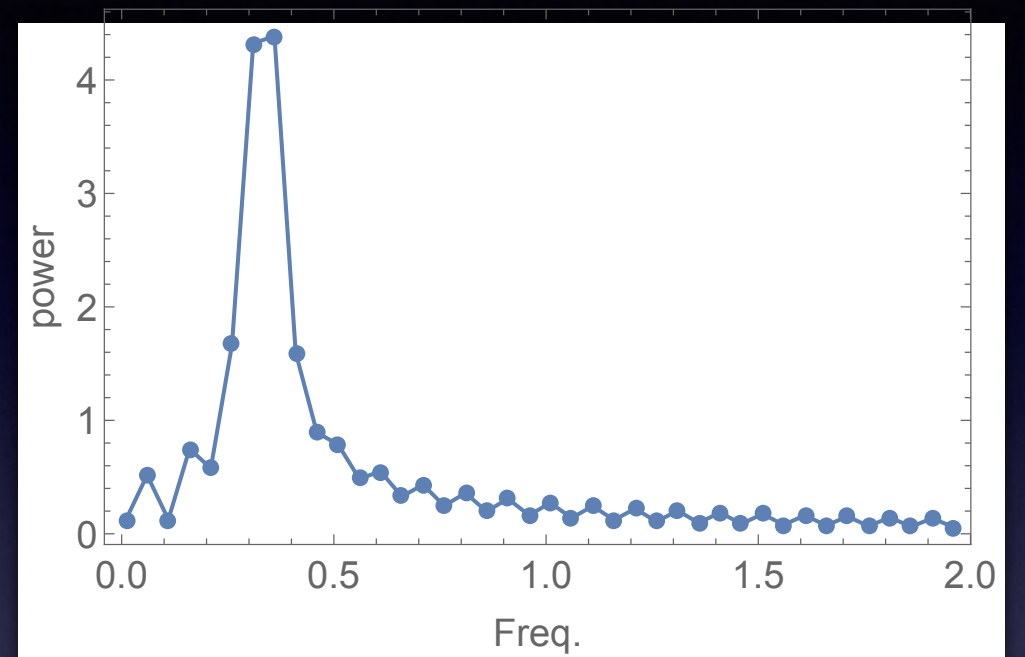
same information in 2d
(position-scale space)



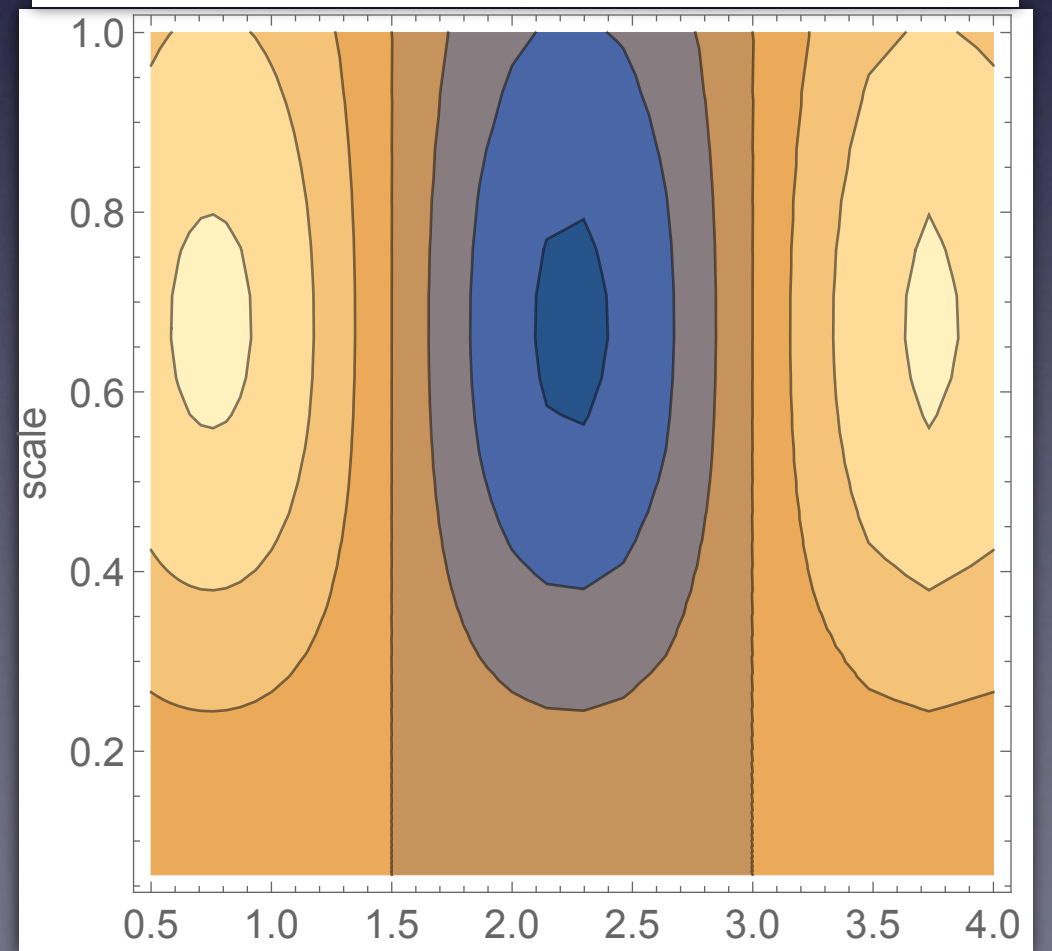
sine wave



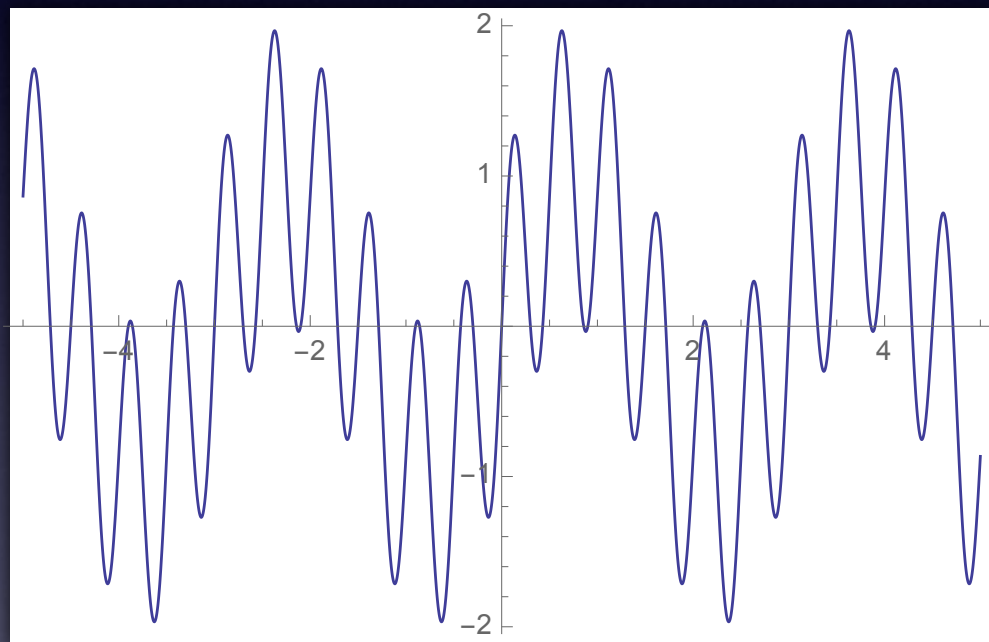
Fourier



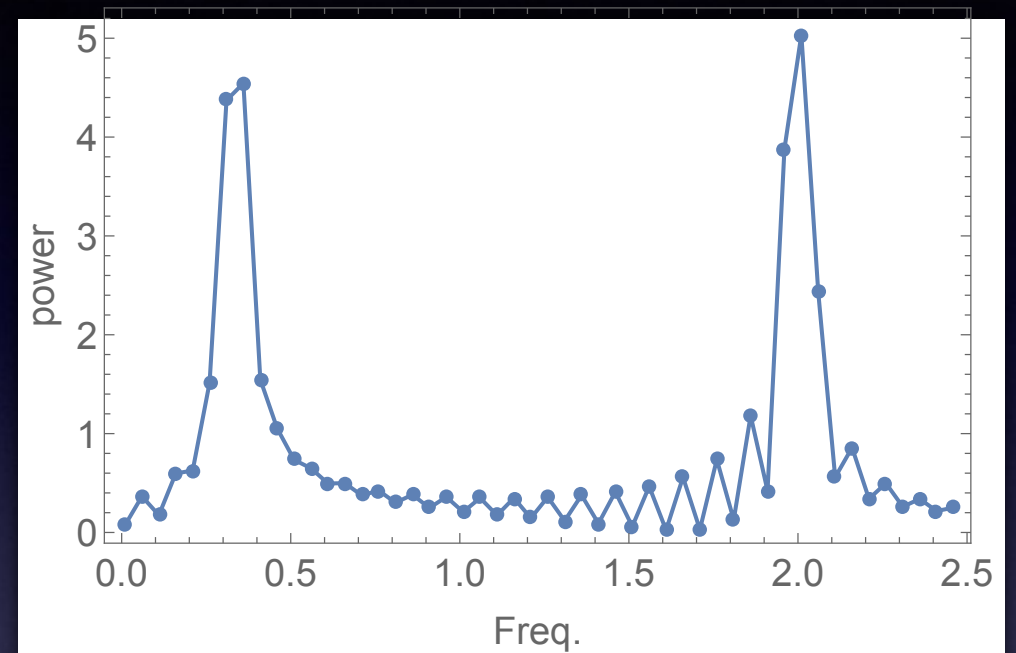
wavelet
Mex. hat



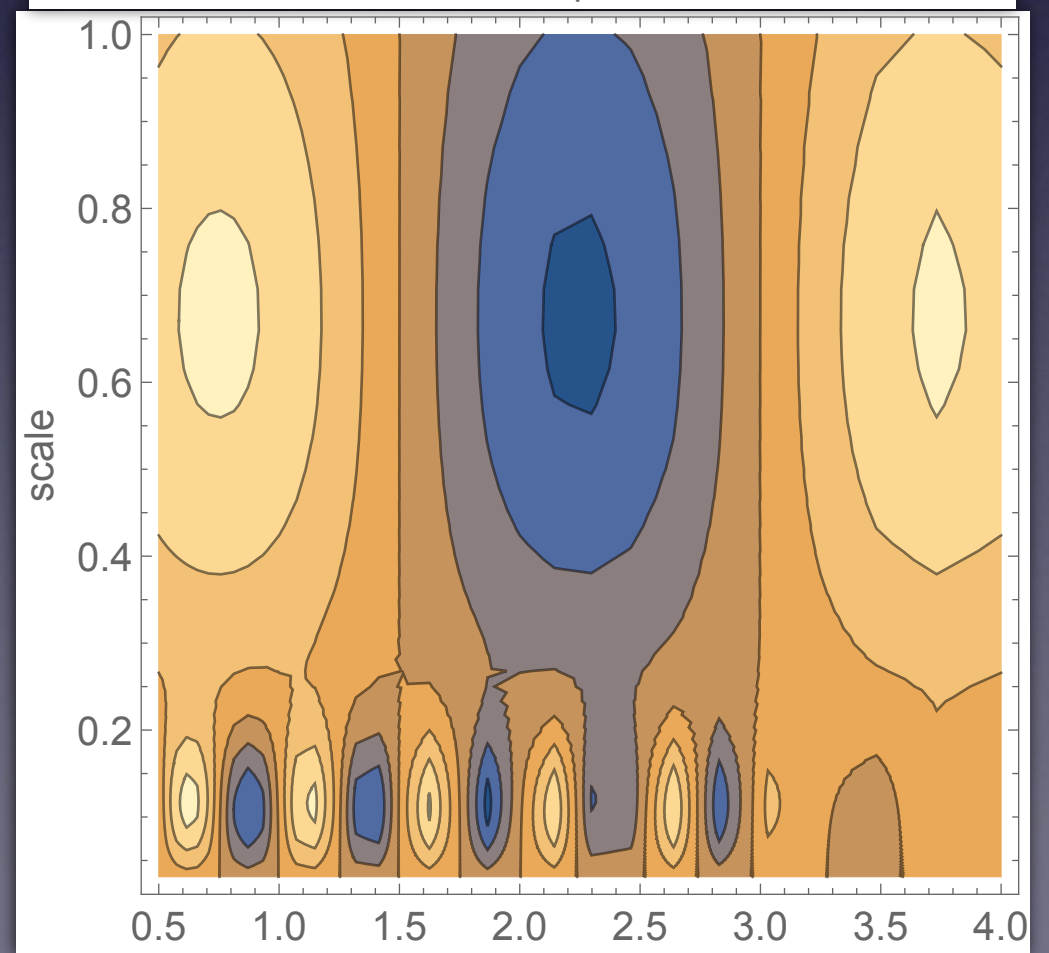
two sine waves



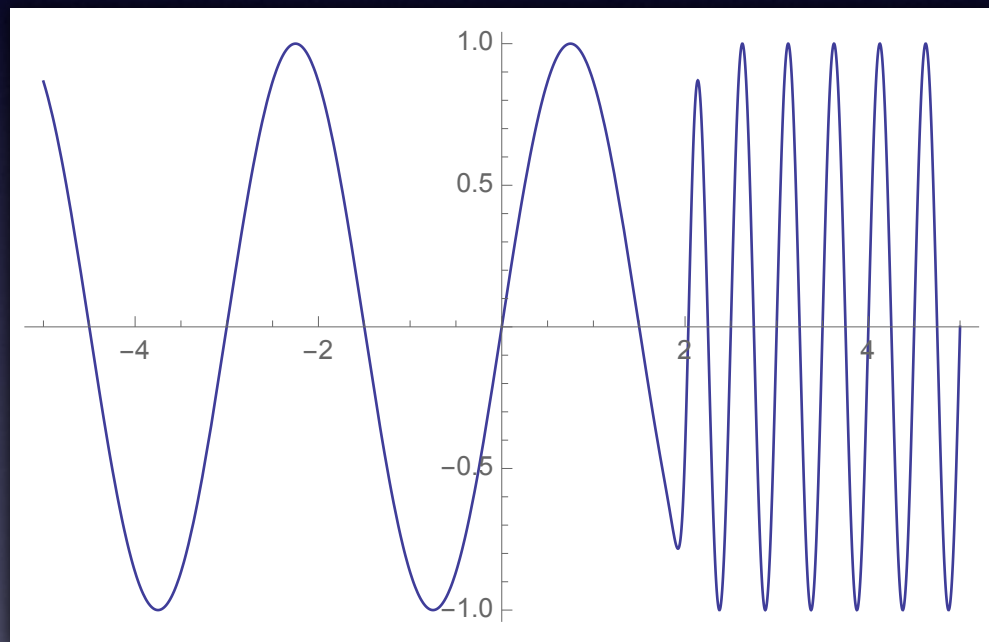
Fourier



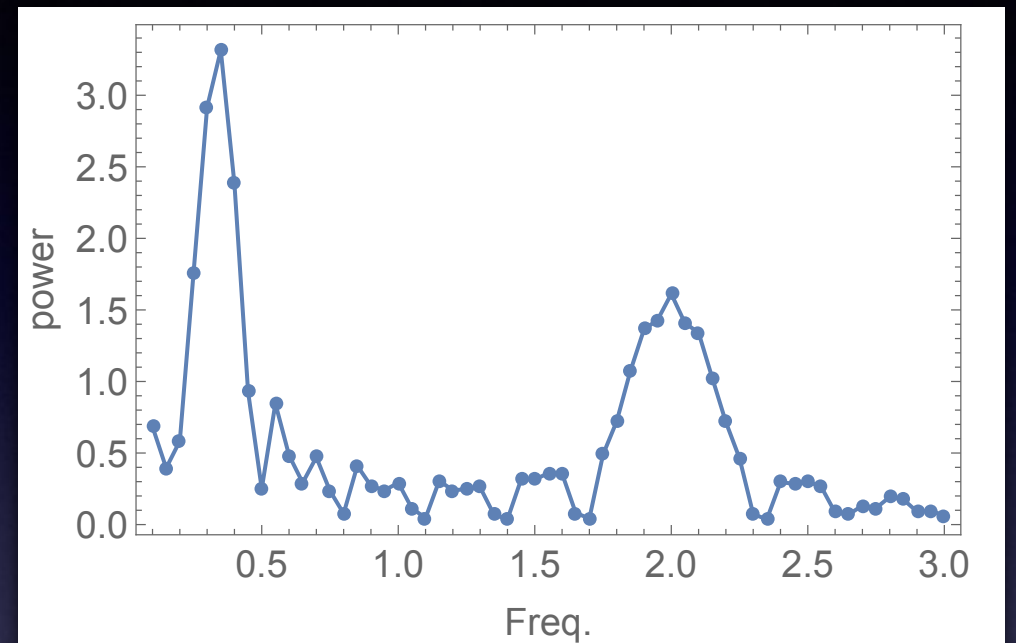
wavelet
Mex. hat



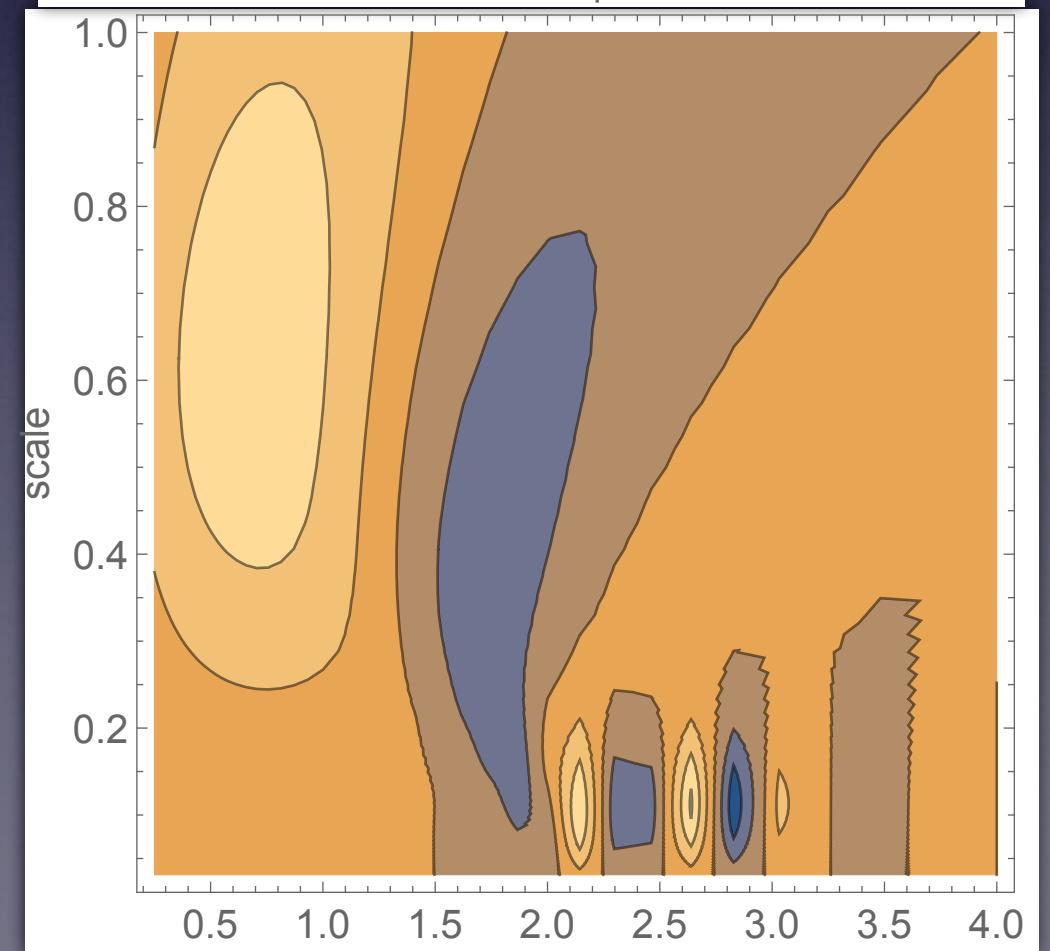
sine waves with transition



Fourier



wavelet
Mex. hat



How might this approach improve upon templates?

GeV sky can be thought of as a high resolution picture; wavelets can find structures in it

Poisson noise and SM uncertainty dominate at scales that are small relative to bubbles or NFW, and the wavelets can identify those scales

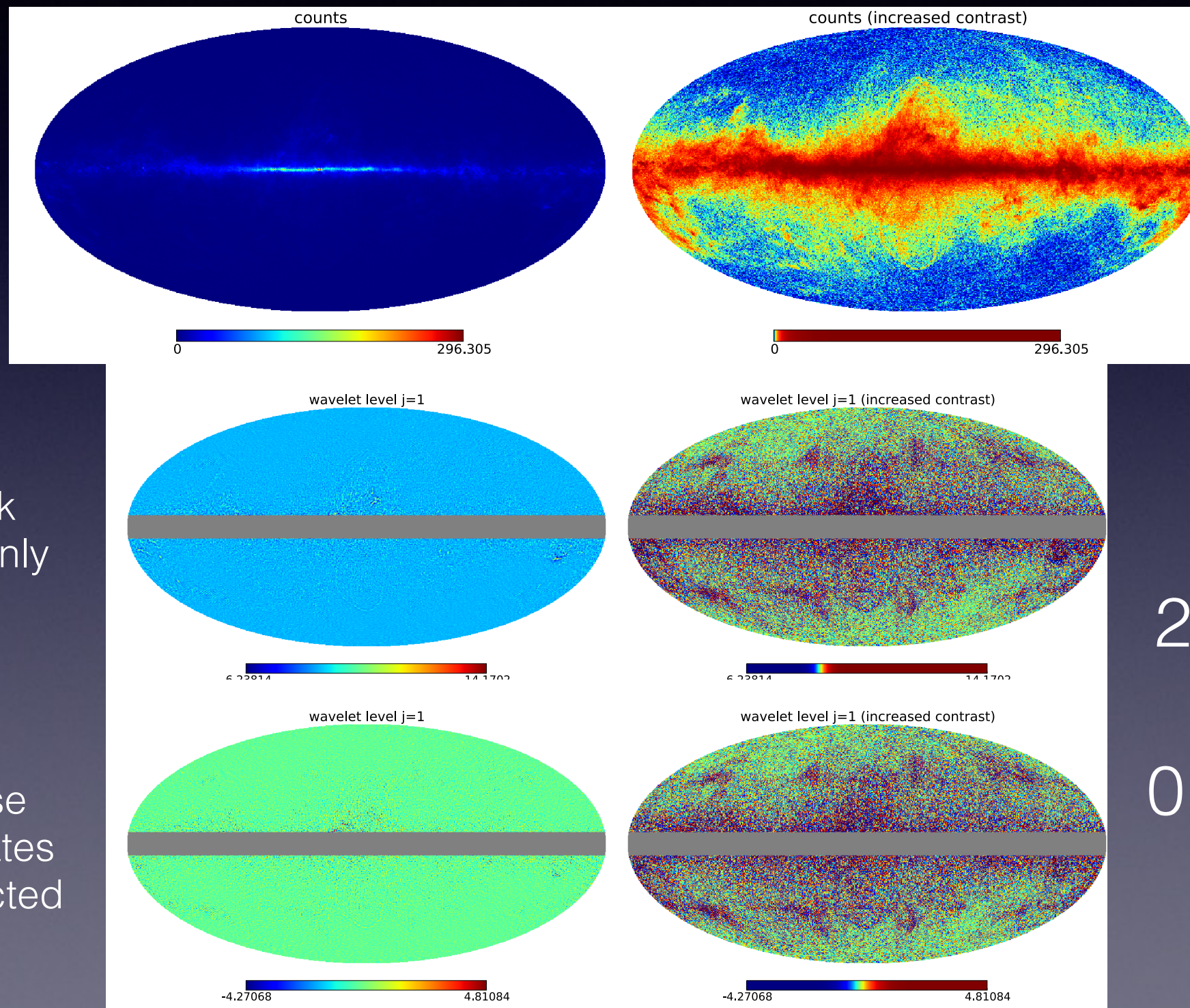
How might this approach improve upon templates?

GeV sky can be thought of as a high resolution picture; wavelets can find structures in it

Poisson noise and SM uncertainty dominate at scales that are small relative to bubbles or NFW, and the wavelets can identify those scales

by identifying and removing such structures, wavelets provide a background expectation that is (relatively) robust against systematic astrophysics uncertainties

Example (mock data)



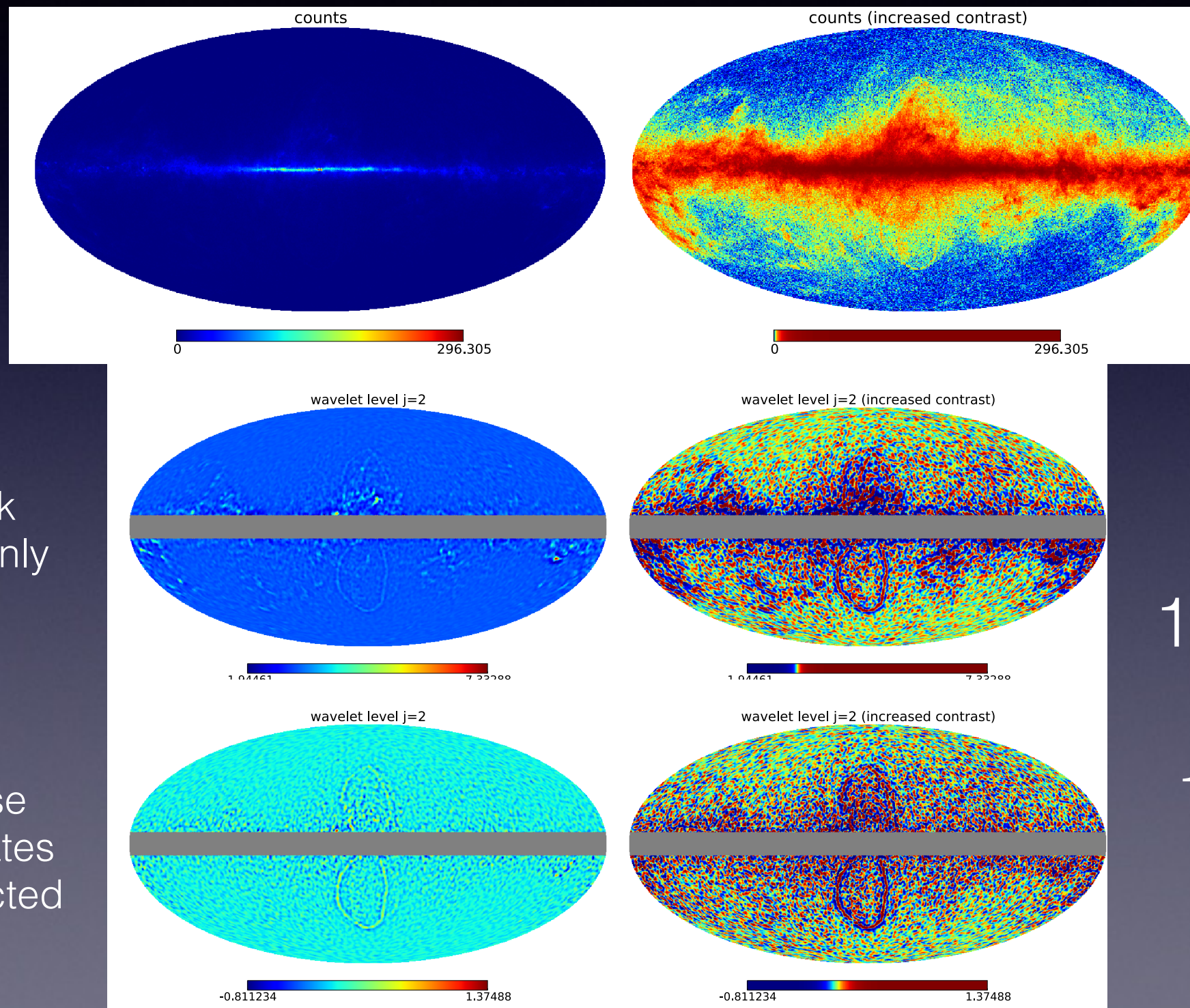
$$\ell_{\max}=512$$

$$256 < \ell < 512$$



$$0.7^\circ < \theta < 1.4^\circ$$

Example (mock data)



$$\ell_{\max}=512$$

mock
data only

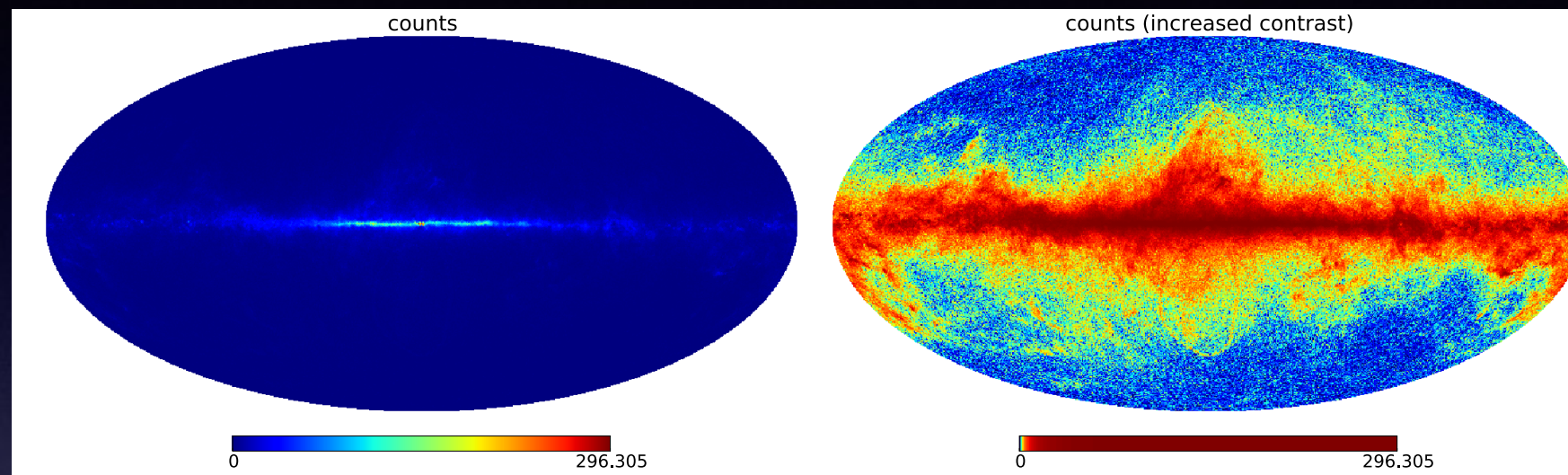
$$128 < \ell < 256$$



$$1.4^\circ < \theta < 3^\circ$$

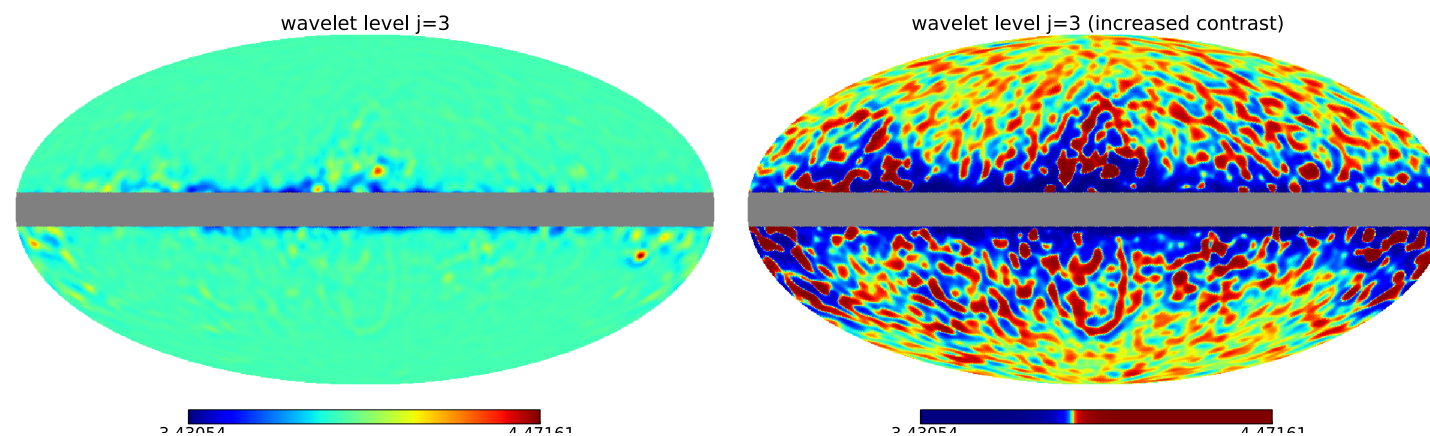
diffuse
templates
subtracted

Example (mock data)



$$\ell_{\max}=512$$

mock
data only

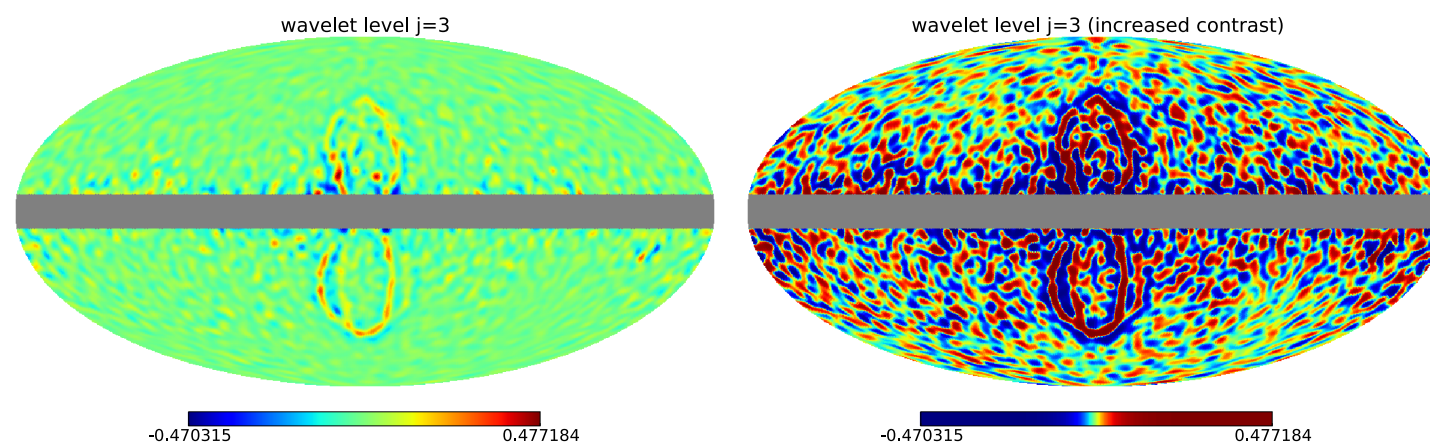


$$64 < \ell < 128$$

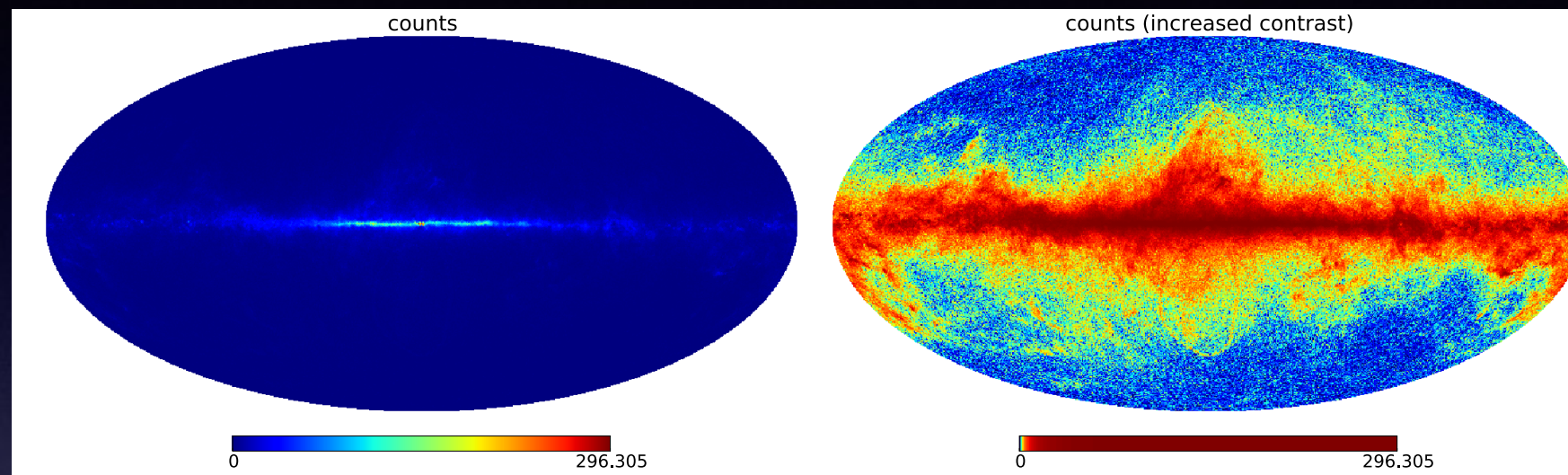


$$3^\circ < \theta < 6^\circ$$

diffuse
templates
subtracted

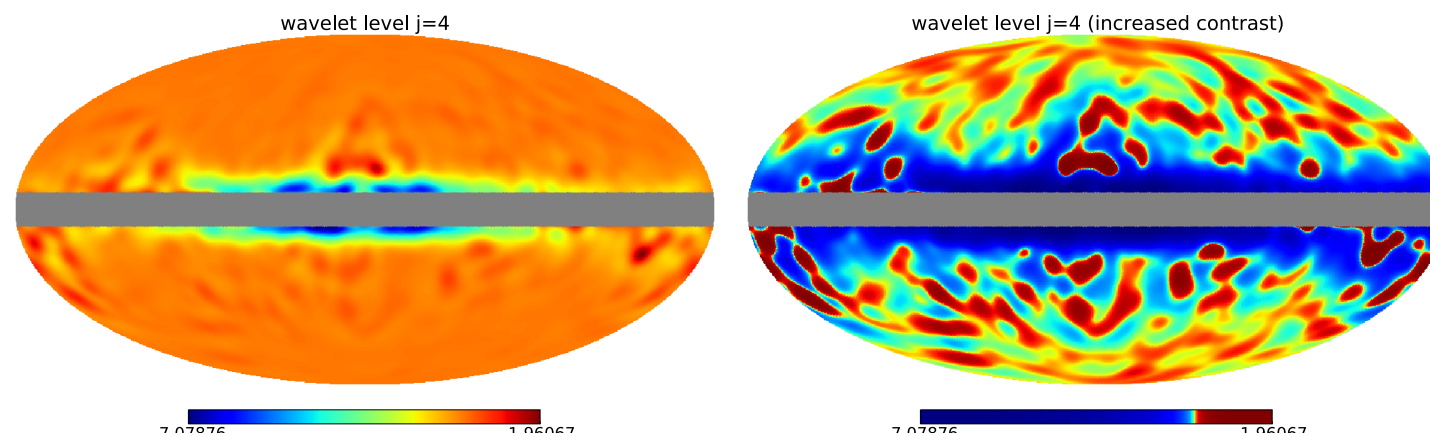


Example (mock data)

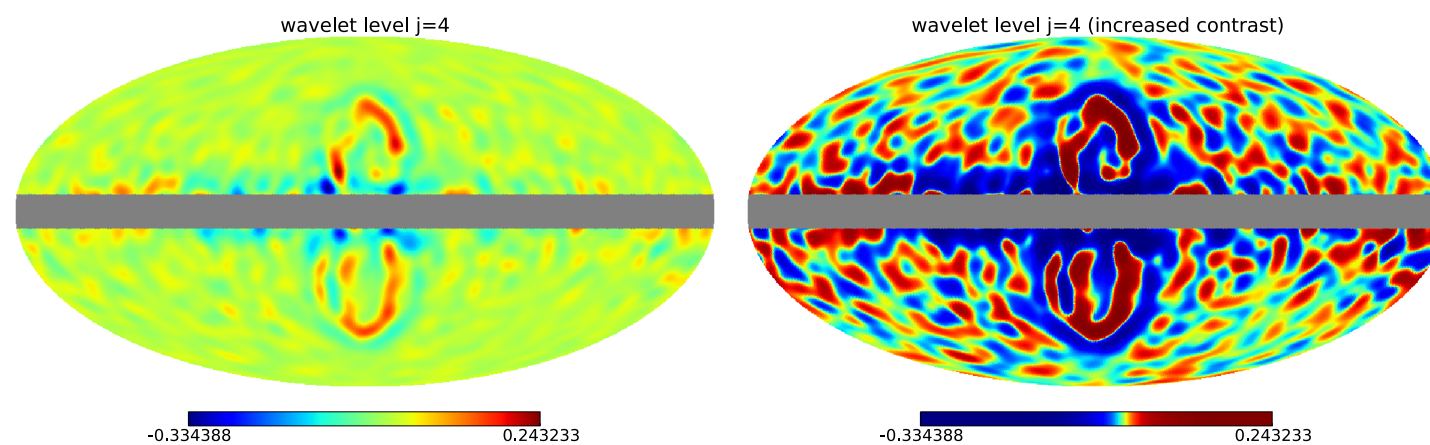


$$\ell_{\max}=512$$

mock
data only



diffuse
templates
subtracted

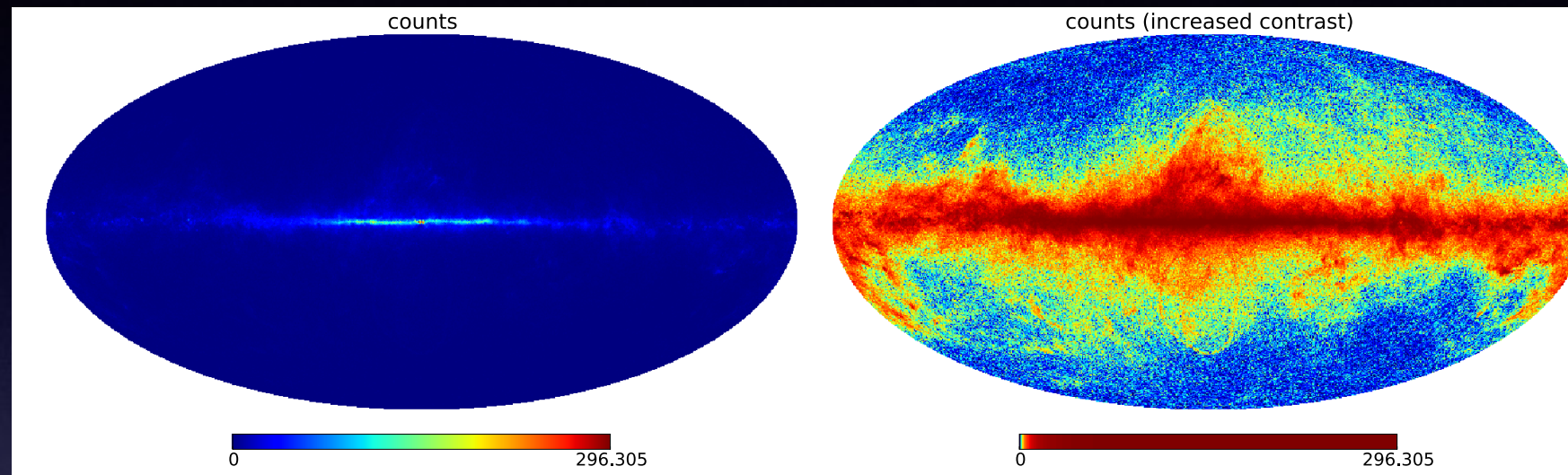


$$32 < \ell < 64$$



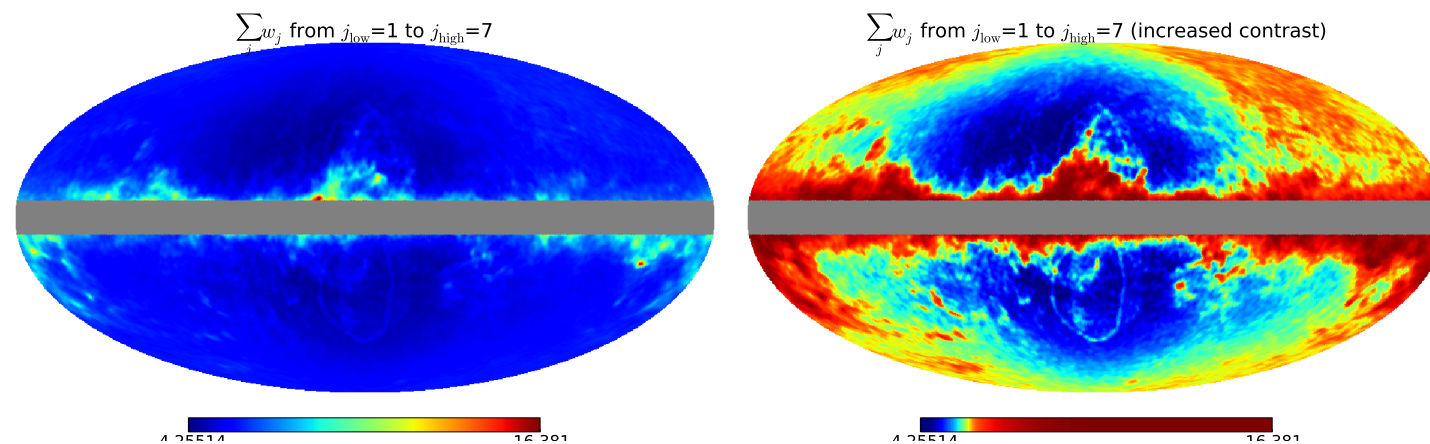
$$6^\circ < \theta < 10^\circ$$

Example (mock data)

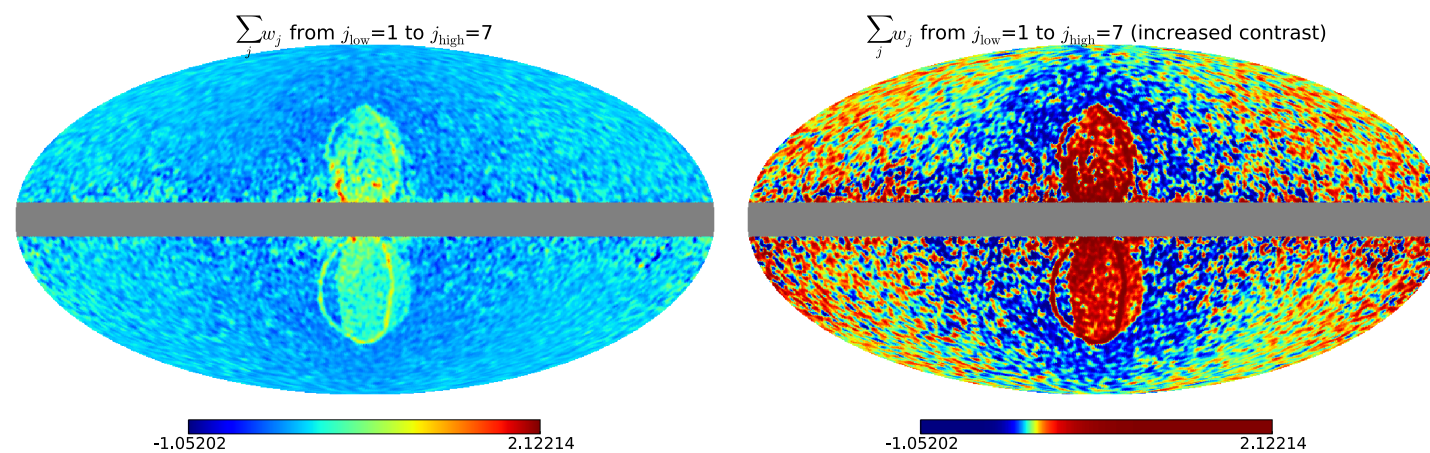


$$\ell_{\max}=512$$

mock
data only



diffuse
templates
subtracted

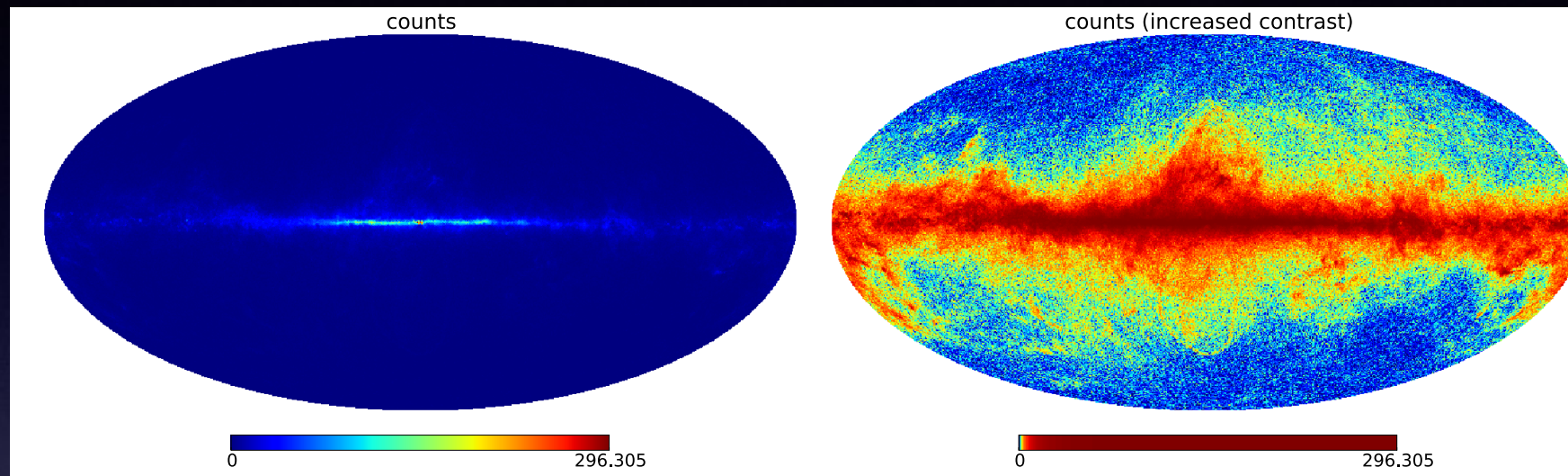


$$4 < \ell < 256$$



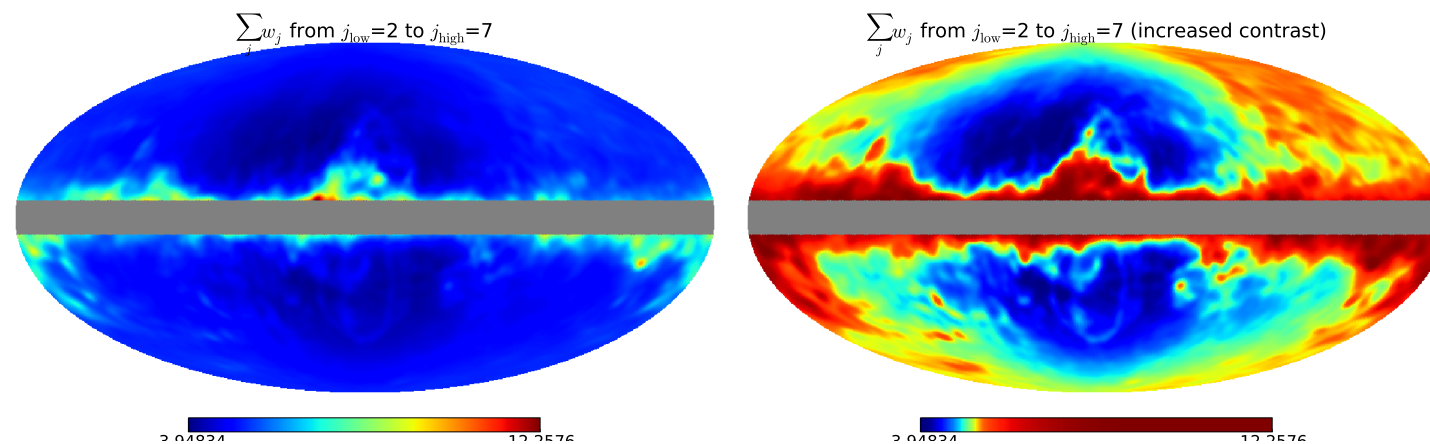
$$1.4^\circ < \theta < 90^\circ$$

Example (mock data)

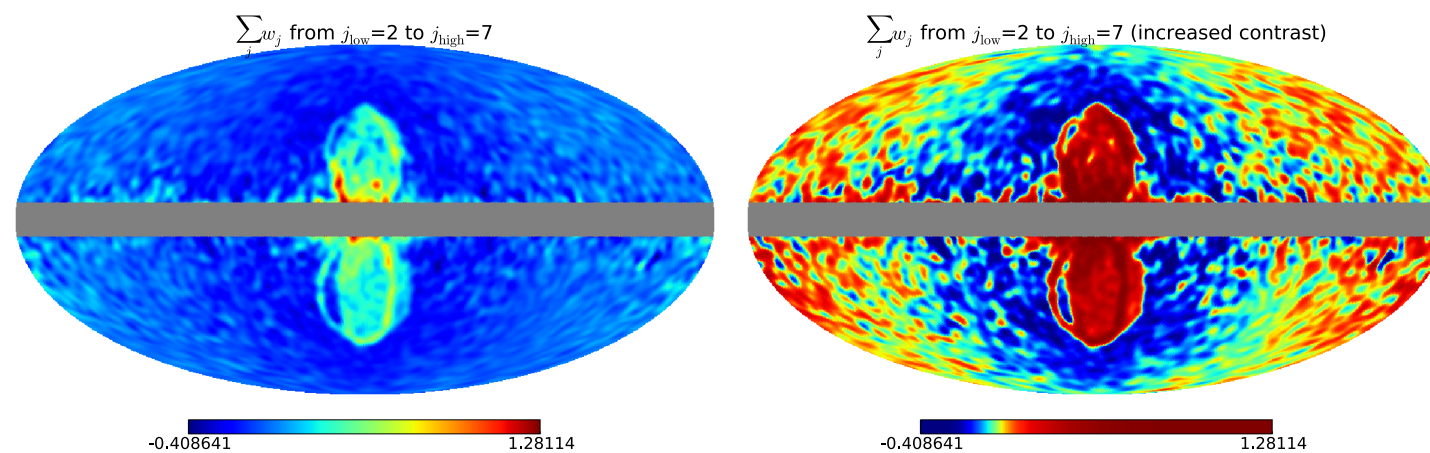


$$\ell_{\max}=512$$

mock
data only



diffuse
templates
subtracted

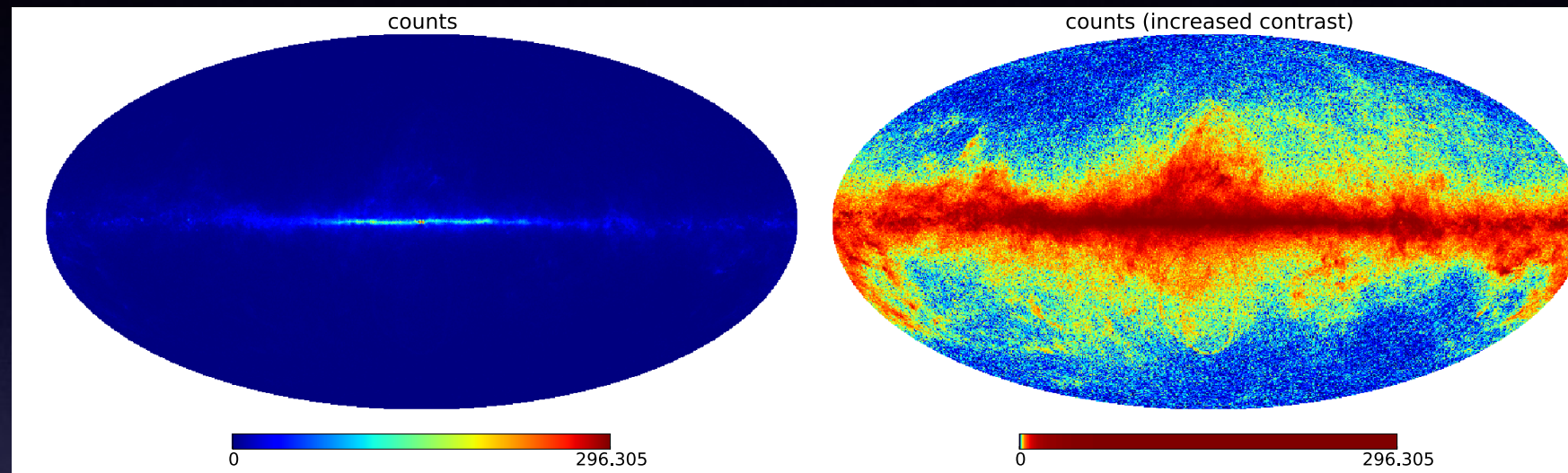


$$4 < \ell < 128$$



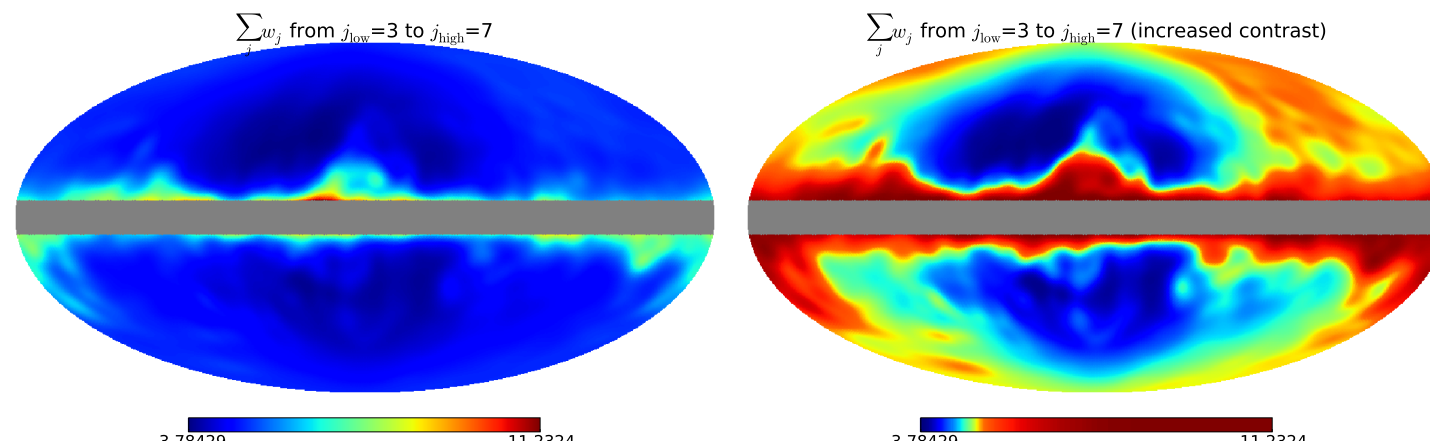
$$3^\circ < \theta < 90^\circ$$

Example (mock data)

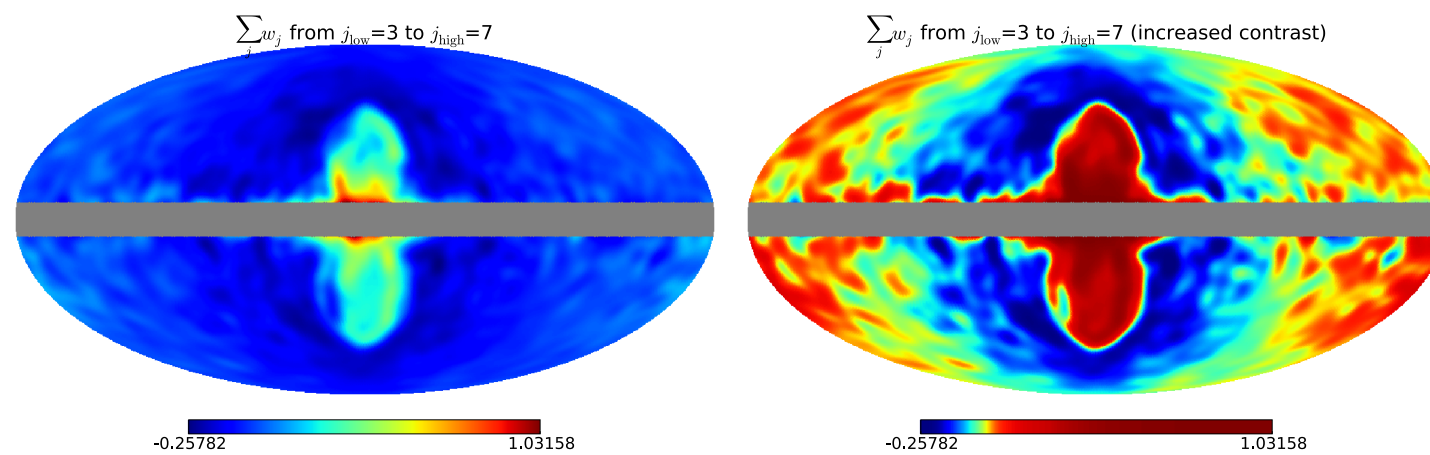


$$\ell_{\max}=512$$

mock
data only



diffuse
templates
subtracted

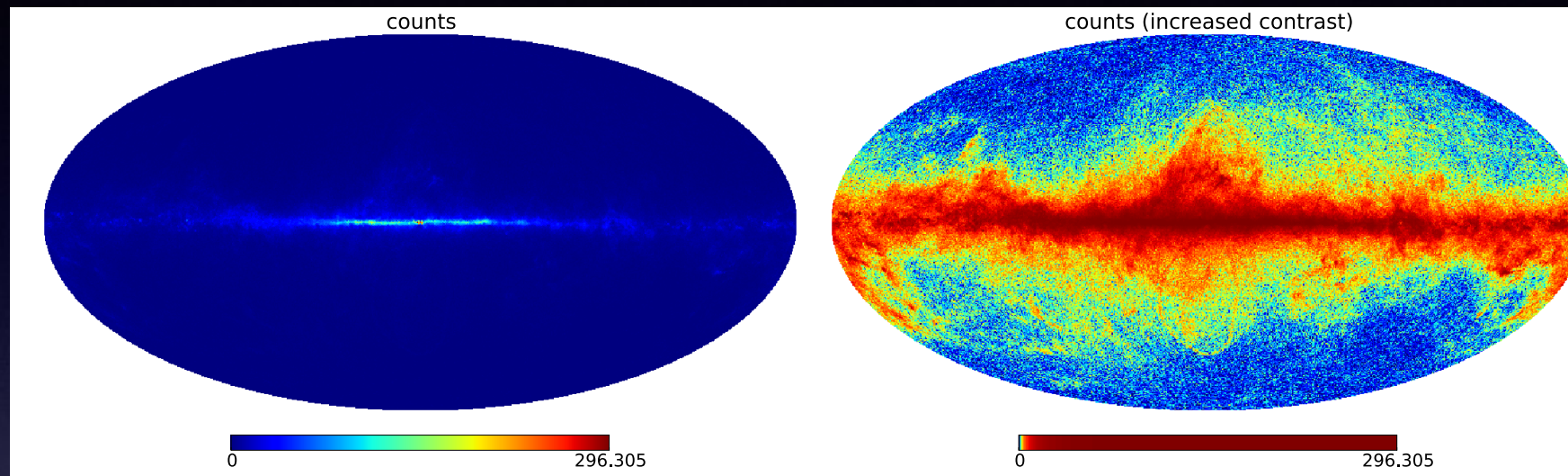


$$4 < \ell < 64$$



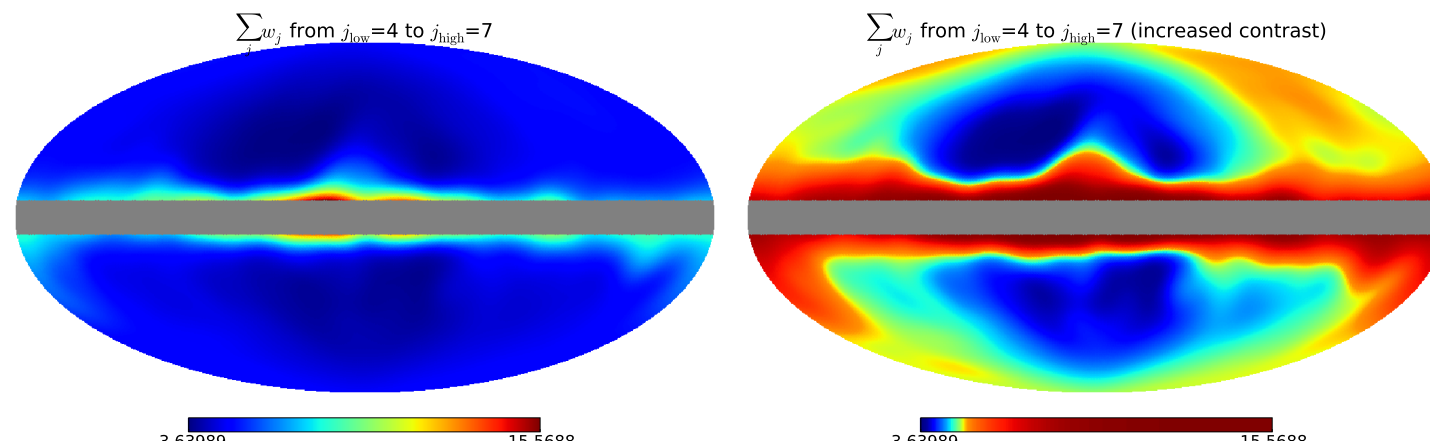
$$6^\circ < \theta < 90^\circ$$

Example (mock data)

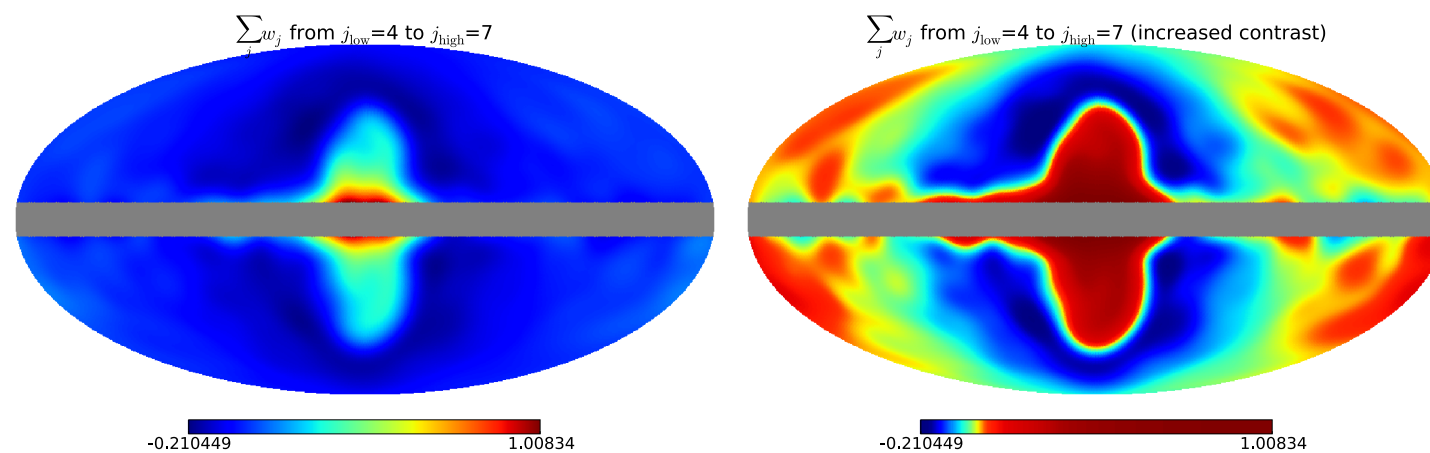


$$\ell_{\max}=512$$

mock
data only



diffuse
templates
subtracted

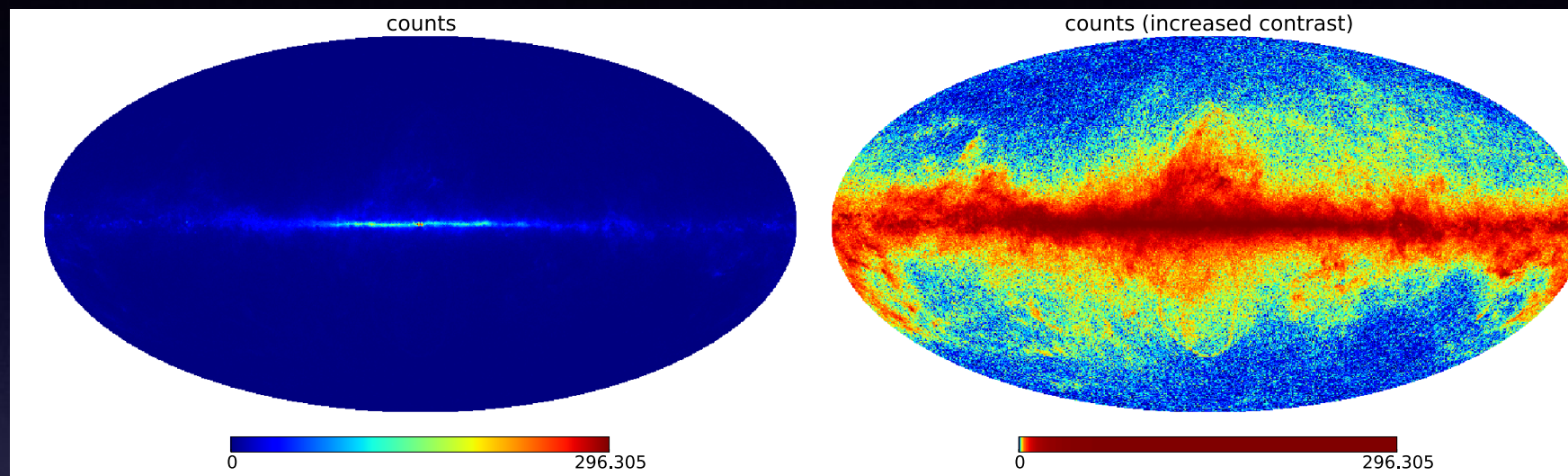


$$4 < \ell < 32$$



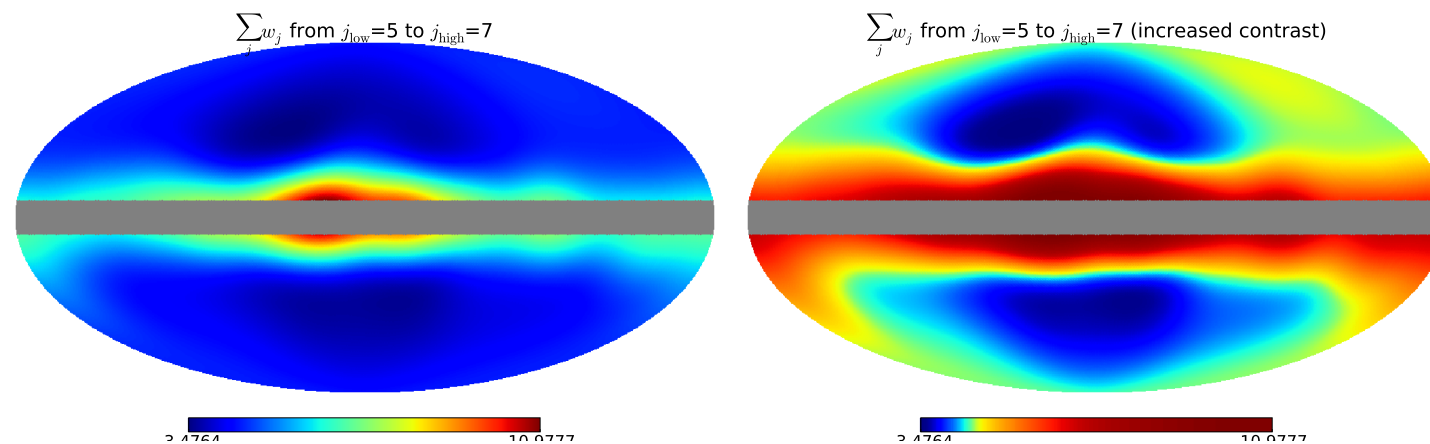
$$10^\circ < \theta < 90^\circ$$

Example (mock data)

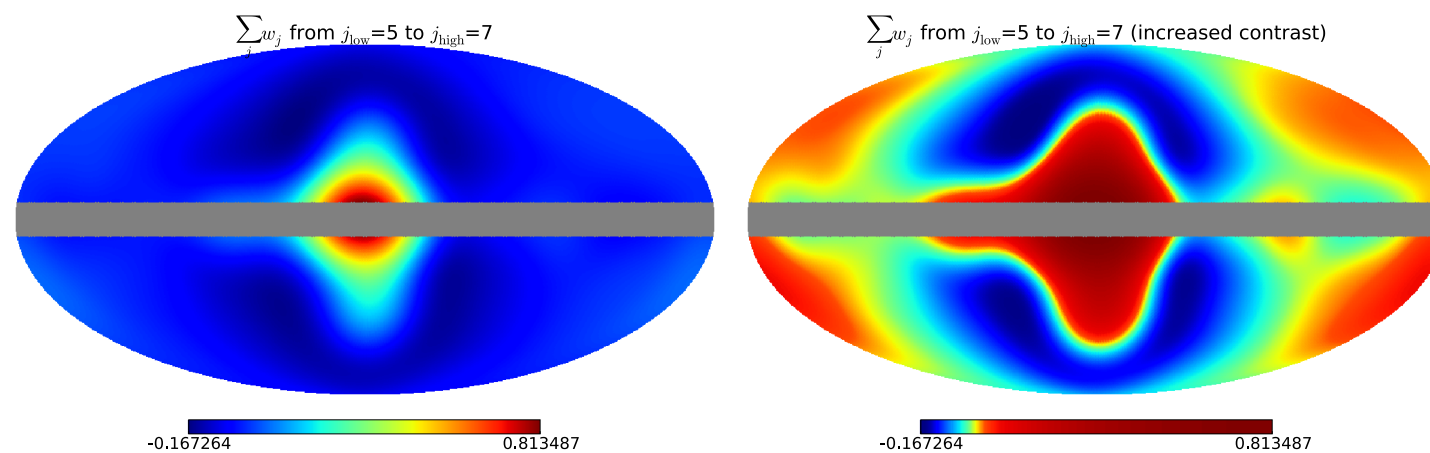


$$\ell_{\max}=512$$

mock
data only



diffuse
templates
subtracted



$$4 < \ell < 16$$



$$22^\circ < \theta < 90^\circ$$

Lesson:

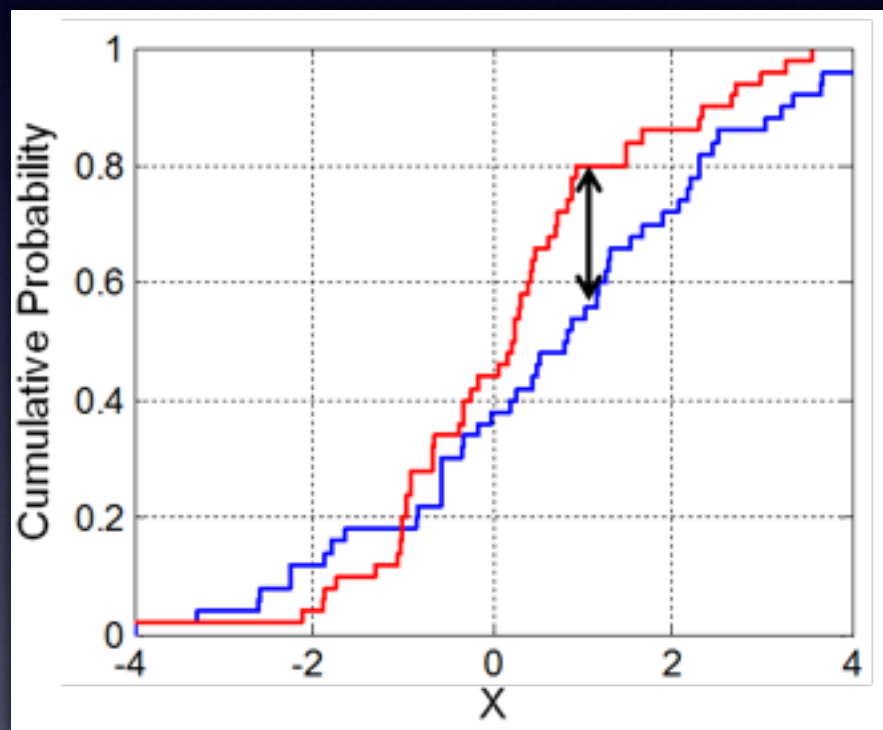
Getting rid of some wavelet levels can provide a much clearer picture of a signal

Question:

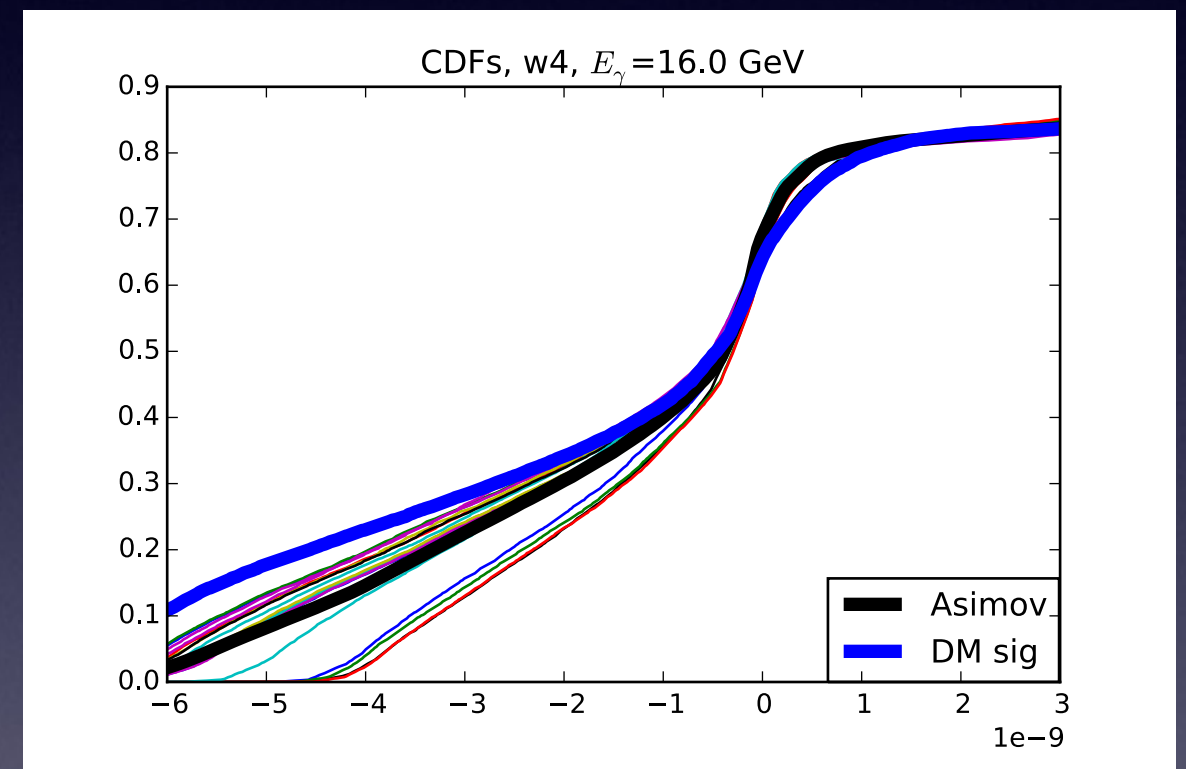
How can we do this in a data-driven (model-independent) (unbiased) (etc....) way?

Kolmogorov-Smirnov Test

maximum distance between two CDFs

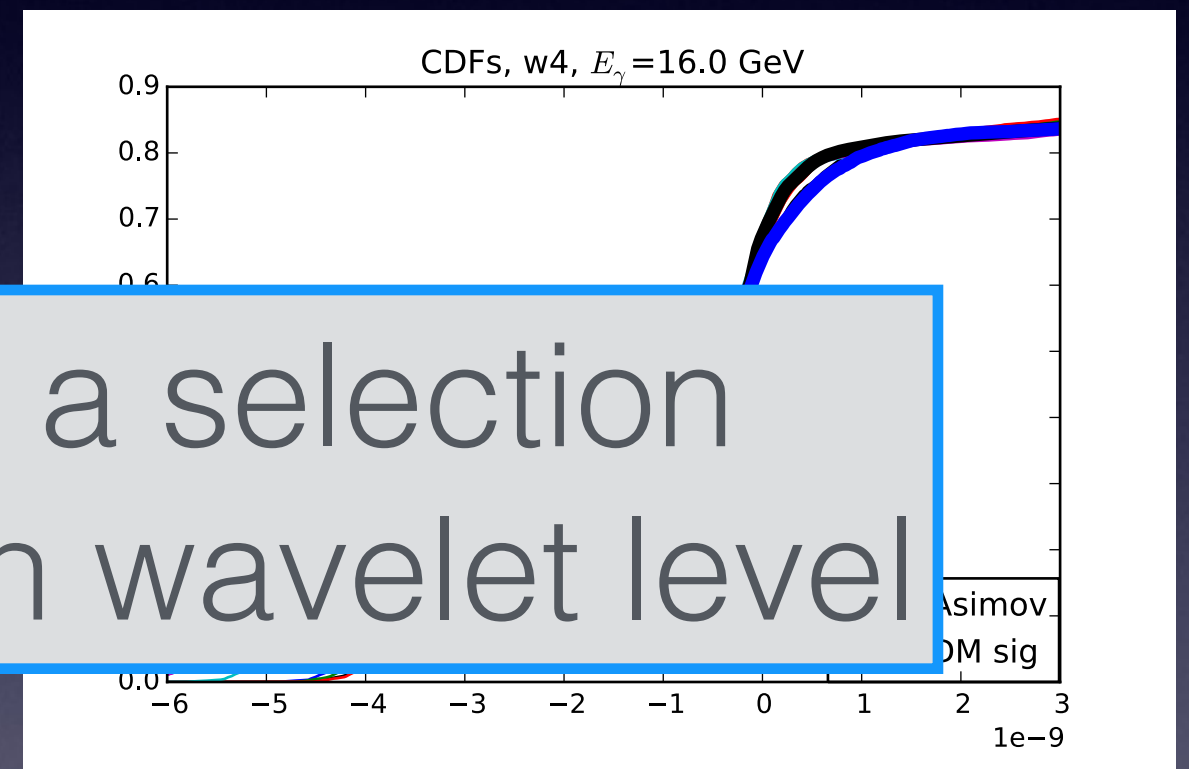
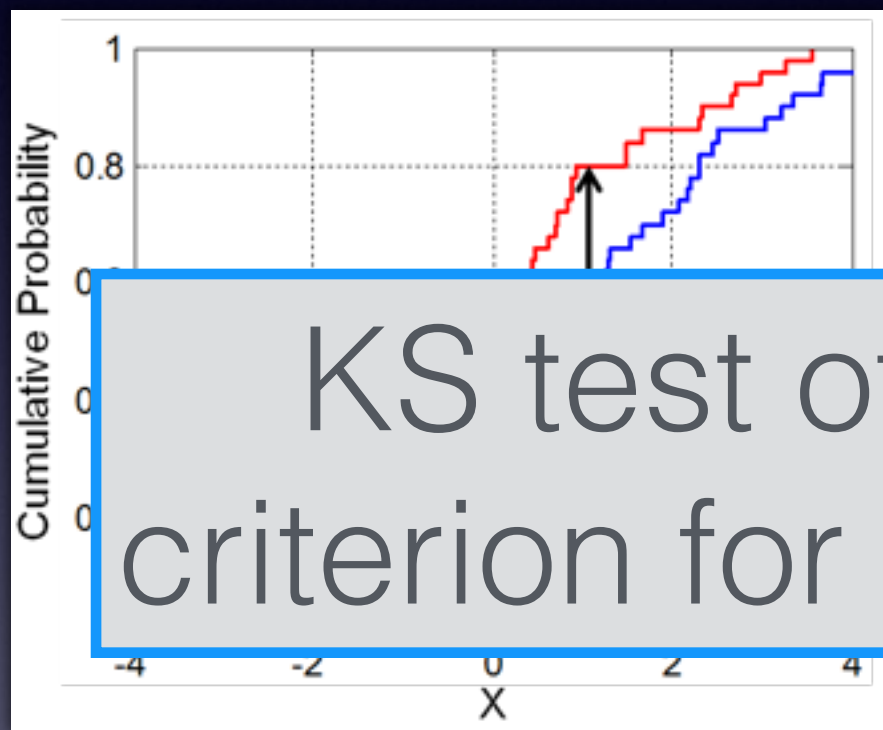


"KS2 Example" by Bscan - Own work.
Licensed under CC0 via Commons -
[https://commons.wikimedia.org/wiki/
File:KS2_Example.png#/media/
File:KS2_Example.png](https://commons.wikimedia.org/wiki/File:KS2_Example.png#/media/File:KS2_Example.png)



Kolmogorov-Smirnov Test

maximum distance between two CDFs



KS test offers a selection criterion for each wavelet level

"KS2 Example" by Bscan - Own work.
Licensed under CC0 via Commons -
https://commons.wikimedia.org/wiki/File:KS2_Example.png#/media/File:KS2_Example.png

“Thresholded” wavelets

signal = S

set of backgrounds = $\{B_i\}$

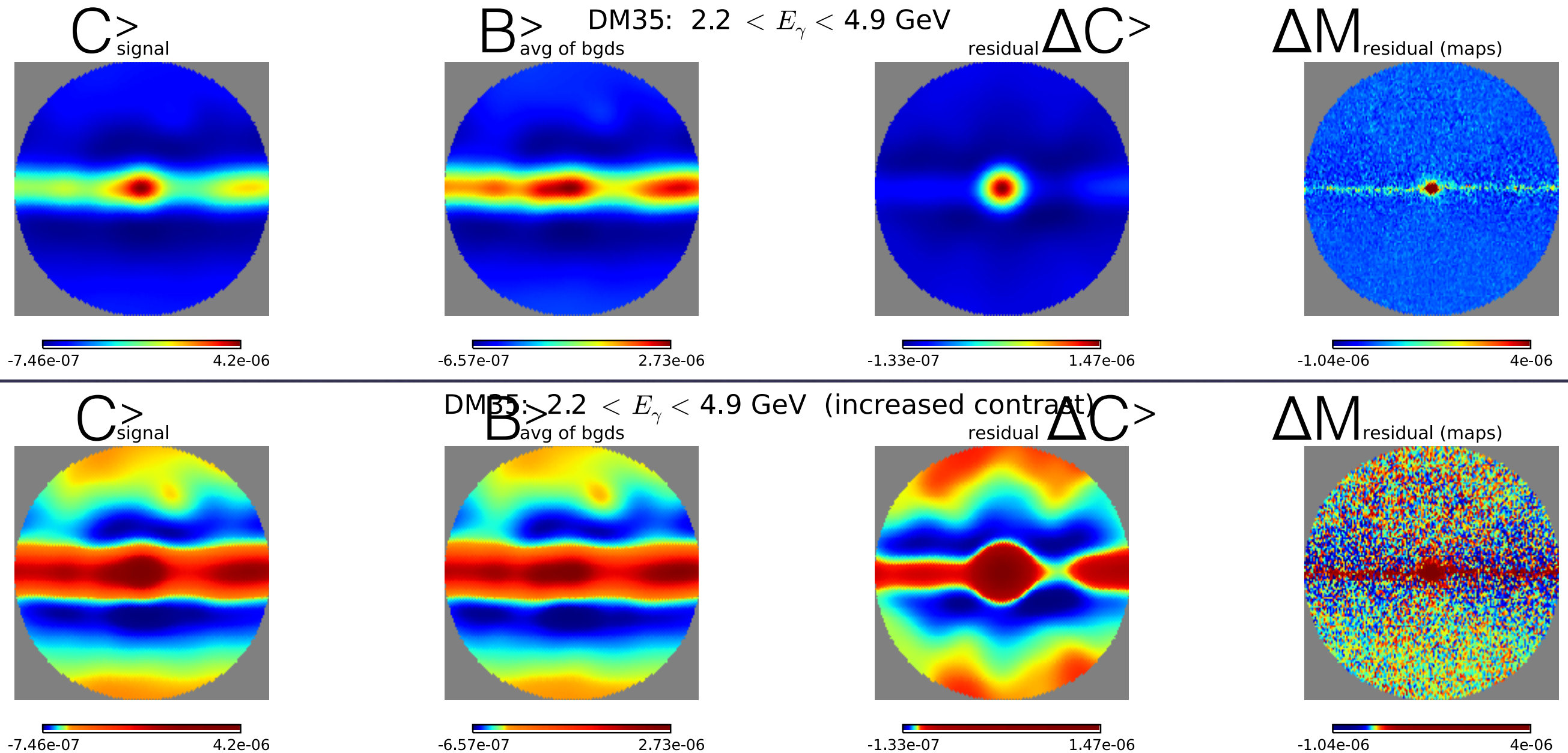
$$w_j^> = \begin{cases} w_j & \text{if } KS(S \mid \text{Asimov}) > 40\% KS(B_i \mid \text{Asimov}) \\ 0 & \text{otherwise} \end{cases}$$

define “cleaned maps:”

$$C^> = \sum_{j=2}^8 w_j^>(S)$$
$$B_i^> = \sum_{j=2}^8 w_j^>(B_i) \Theta[w_j^>(S)]$$
$$B^> = \text{avg}(\{B_i^>\})$$

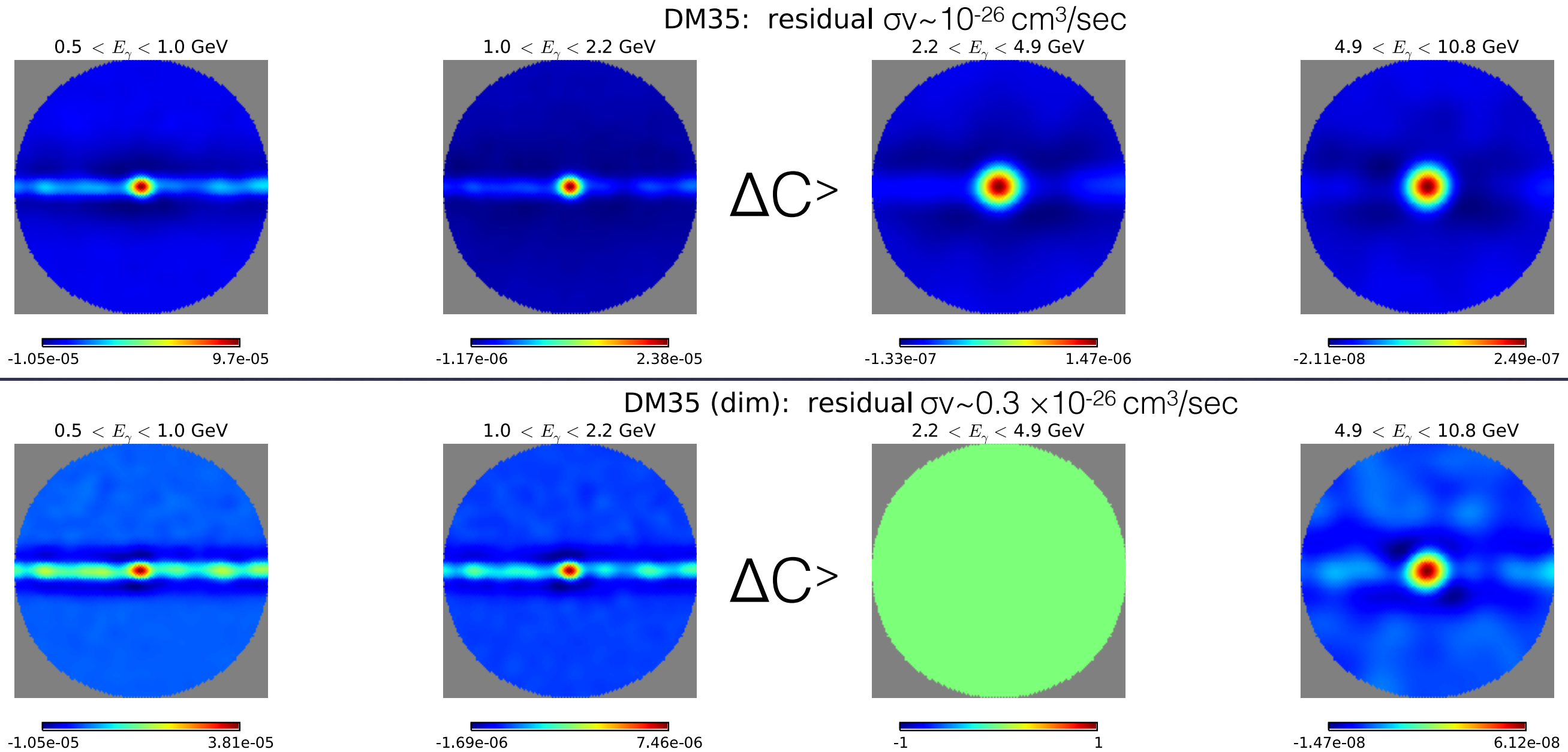
and “cleaned residual:” $\Delta C^> = C^> - B^>$

Cleaned Map Method



wavelets provide clearer residual than maps

Cleaned Map Threshold

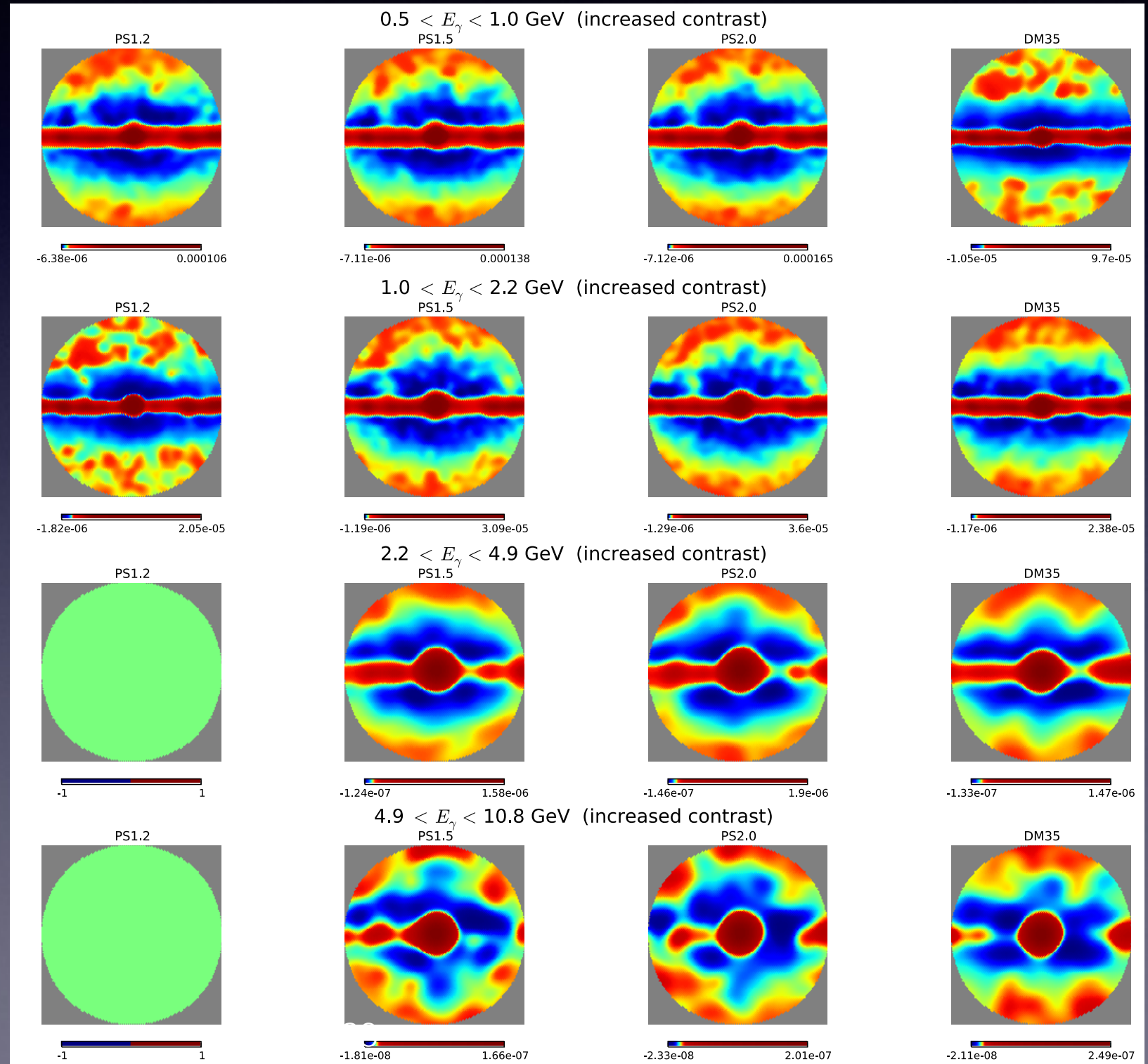


30% as bright is much harder to see

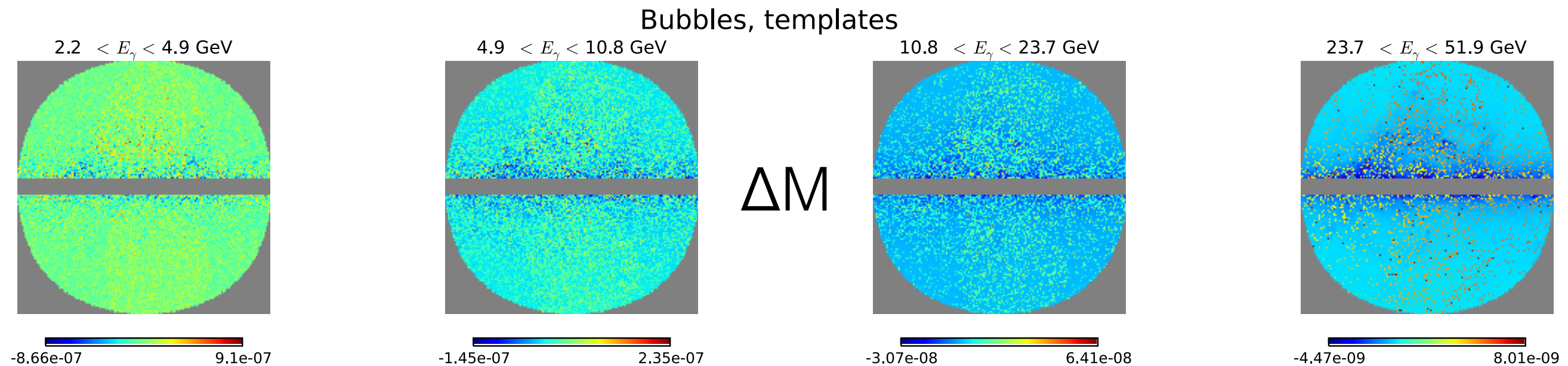
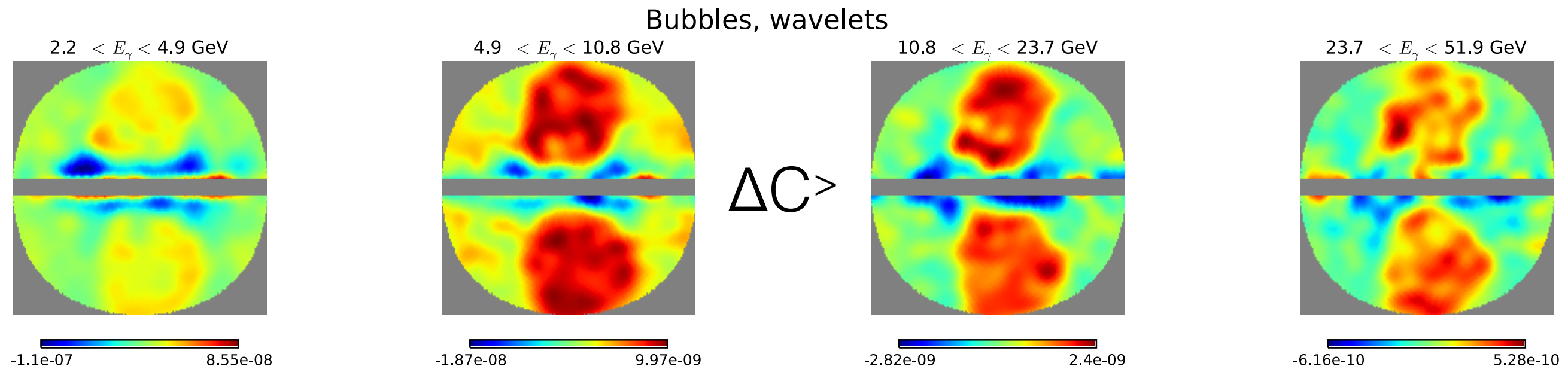
DM vs. Point Sources?

$$L \sim L_0^{-\alpha} e^{-L/L_c}$$

keep $\int L$ fixed



Cleaned Map, Bubbles



What are wavelets?

Allow analysis sensitive to both position and size



different structures have “power” at different levels of the decomposition (edges = sharp variation, important first; larger scale objects = broader variation, important later)



wavelets find structures, and the GCE is a qualitatively new structure that we ought to learn more about

Conclusions

Galactic center gamma ray excess is exciting to follow, but still so much more to learn about it

Need some less-model-dependent information

Wavelets are a promising tool for learning about this data

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Much more to do!

Thanks!